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Title: Beyond Moore's Law: Exploring the Future of Computation

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Beyond Moore's Law

Exploring the Future of Computation



Carleton Coffrin

Advanced Network Science Initiative lanl-ansi.github.io



Who is Carleton?

- Computer Science / Optimization Algorithms
 - Los Alamos National Laboratory
 - Ph.D. Advisor Pascal Van Hentenryck



Math-heuristics



Power System Expertise

- Convex Power Flow Relaxations (QC formulation)
- Power Flow Approximations (LPAC formulation)
- NESTA / PGLib AC-OPF Benchmarking Datasets
- PowerModels.jl (similar to Matpower, in Julia)



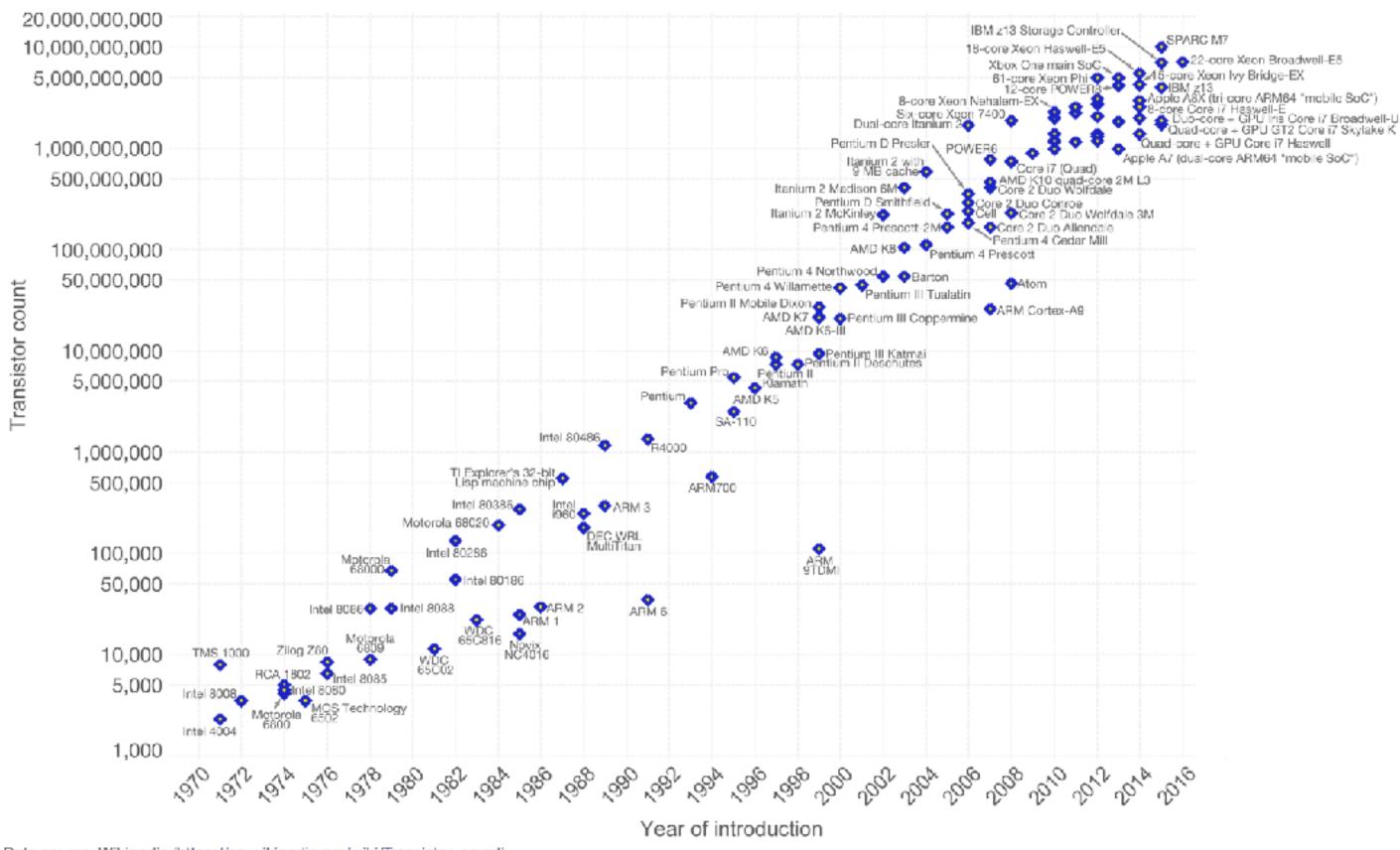
What is Moore's Law

Shrinking transistors have powered 50 years of advances in computing.

Moore's Law – The number of transistors on integrated circuit chips (1971-2016)



Moore's law describes the empirical regularity that the number of transistors on integrated circuits doubles approximately every two years. This advancement is important as other aspects of technological progress – such as processing speed or the price of electronic products – are strongly linked to Moore's law.



Data source: Wikipedia (https://en.wikipedia.org/wiki/Transistor_count)
The data visualization is available at OurWorldinData.org. There you find more visualizations and research on this topic.

Licensed under CC-BY-SA by the author Max Roser.

The Bad News

Intel has suggested silicon transistors can only keep shrinking for another five years. (2016)

Intelligent Machines

Intel: Chips Will Have to Sacrifice Speed Gains for Energy Savings

A major technological shift is needed in the next few years if computer chips are to keep improving.

by Katherine Bourzac February 5, 2016

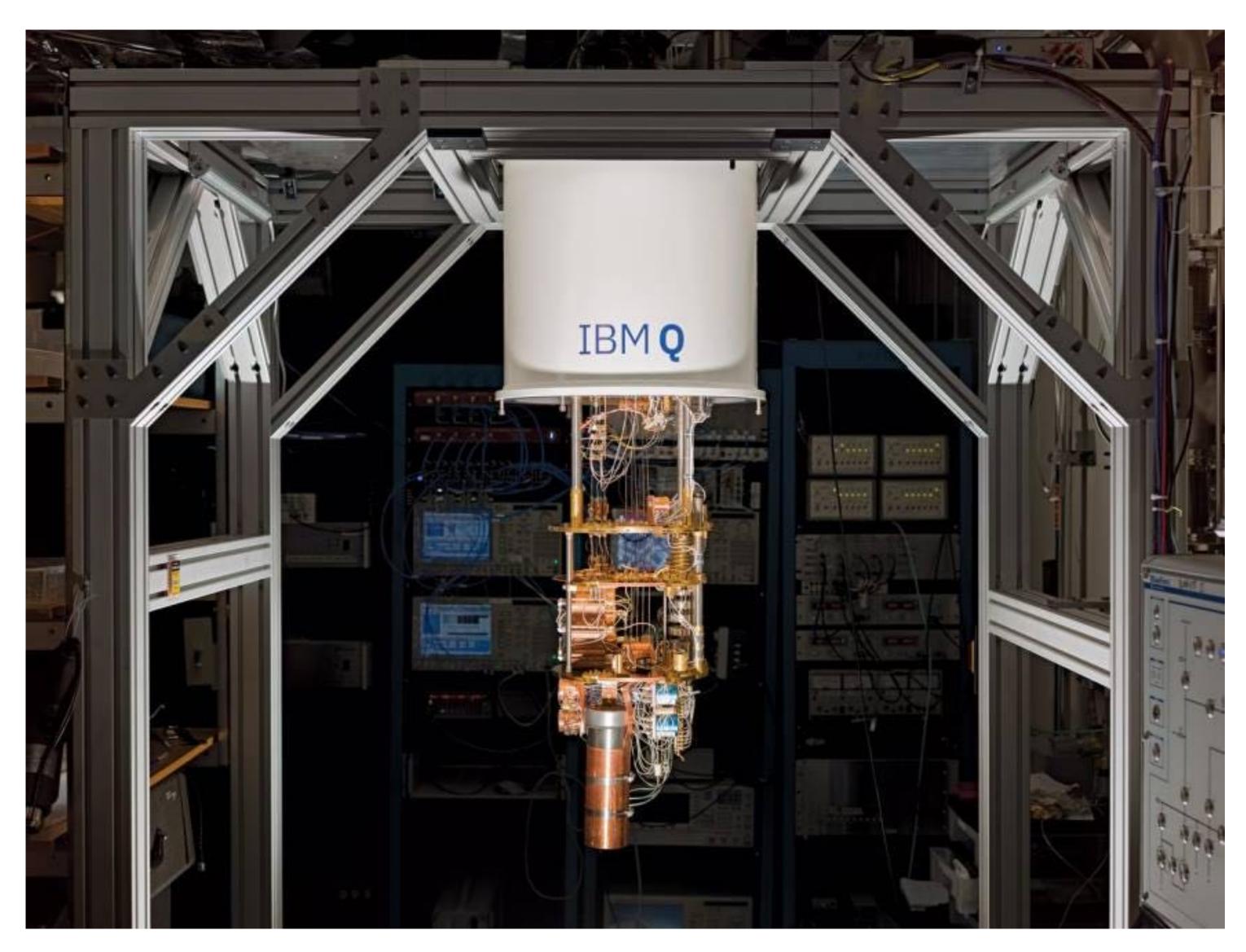
Hitting basic physical limits, e.g. energy/heat per unit area.

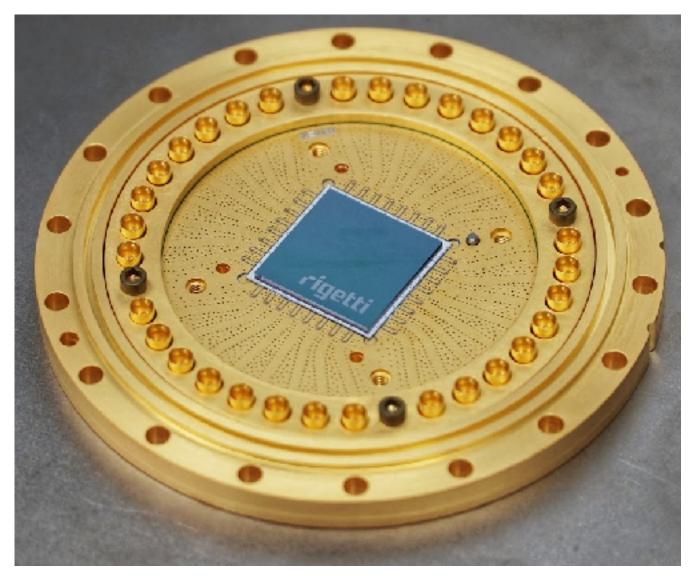
Moore's Law is Dead! Now What?

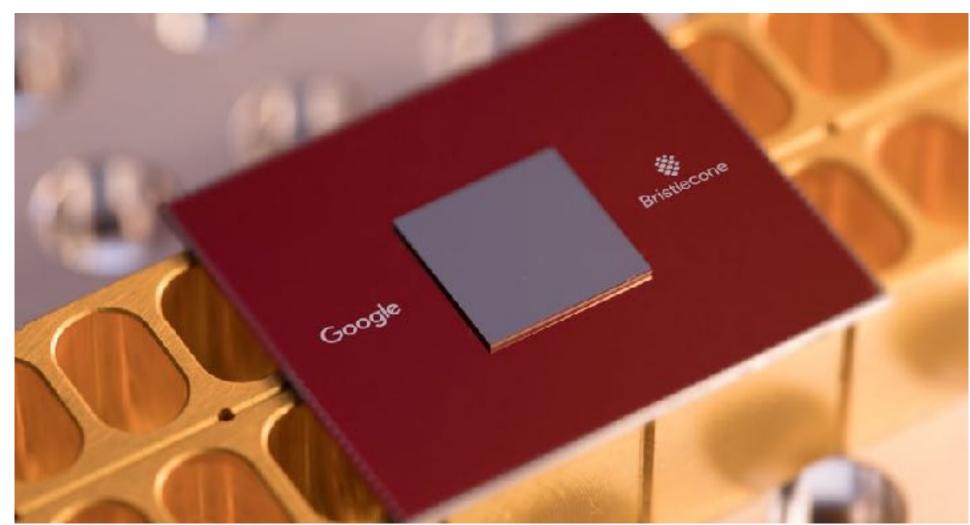
https://www.technologyreview.com/s/601441/moores-law-is-dead-now-what/

https://www.technologyreview.com/s/600716/intel-chips-will-have-to-sacrifice-speed-gains-for-energy-savings/

Universal / Gate-Based Quantium Computers

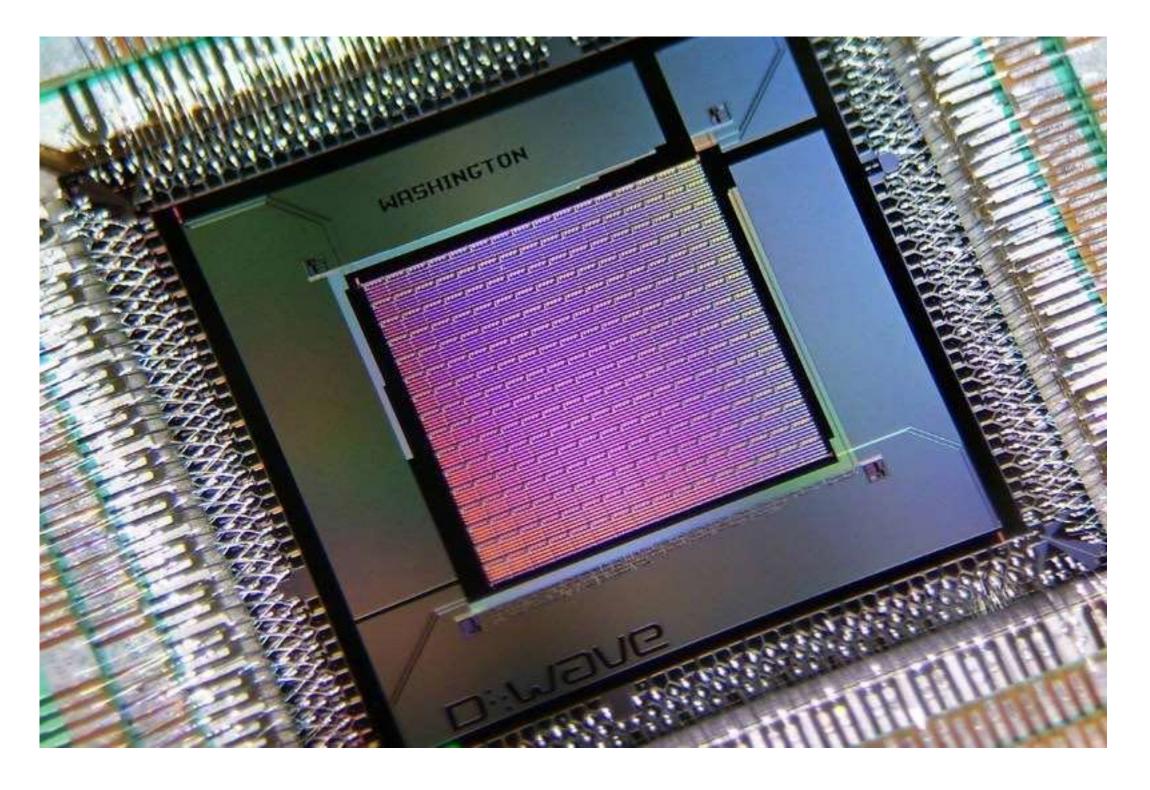






Adiabatic Quantium Computers

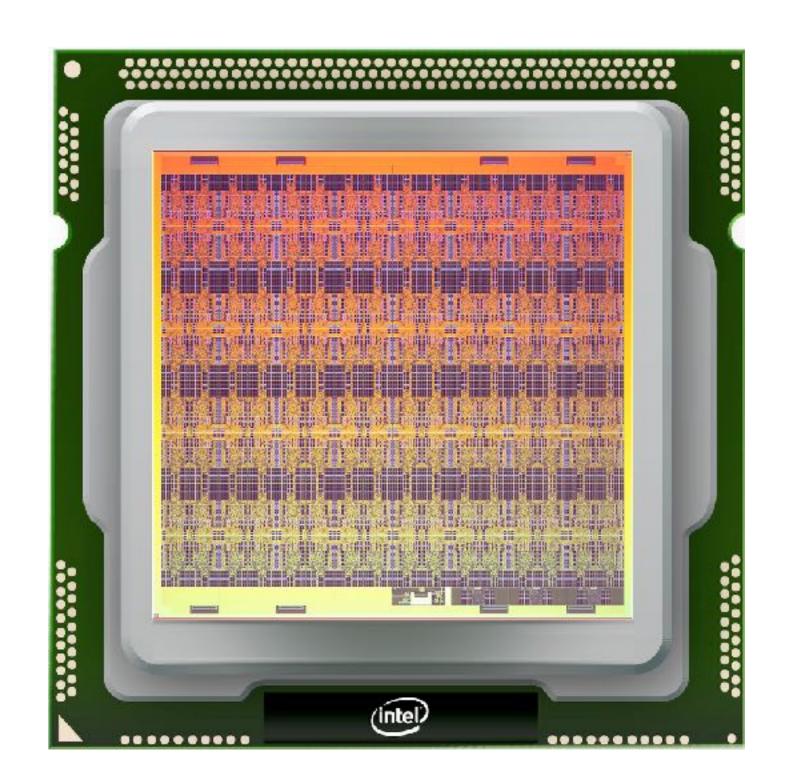




Neuromorphic Coprocessors

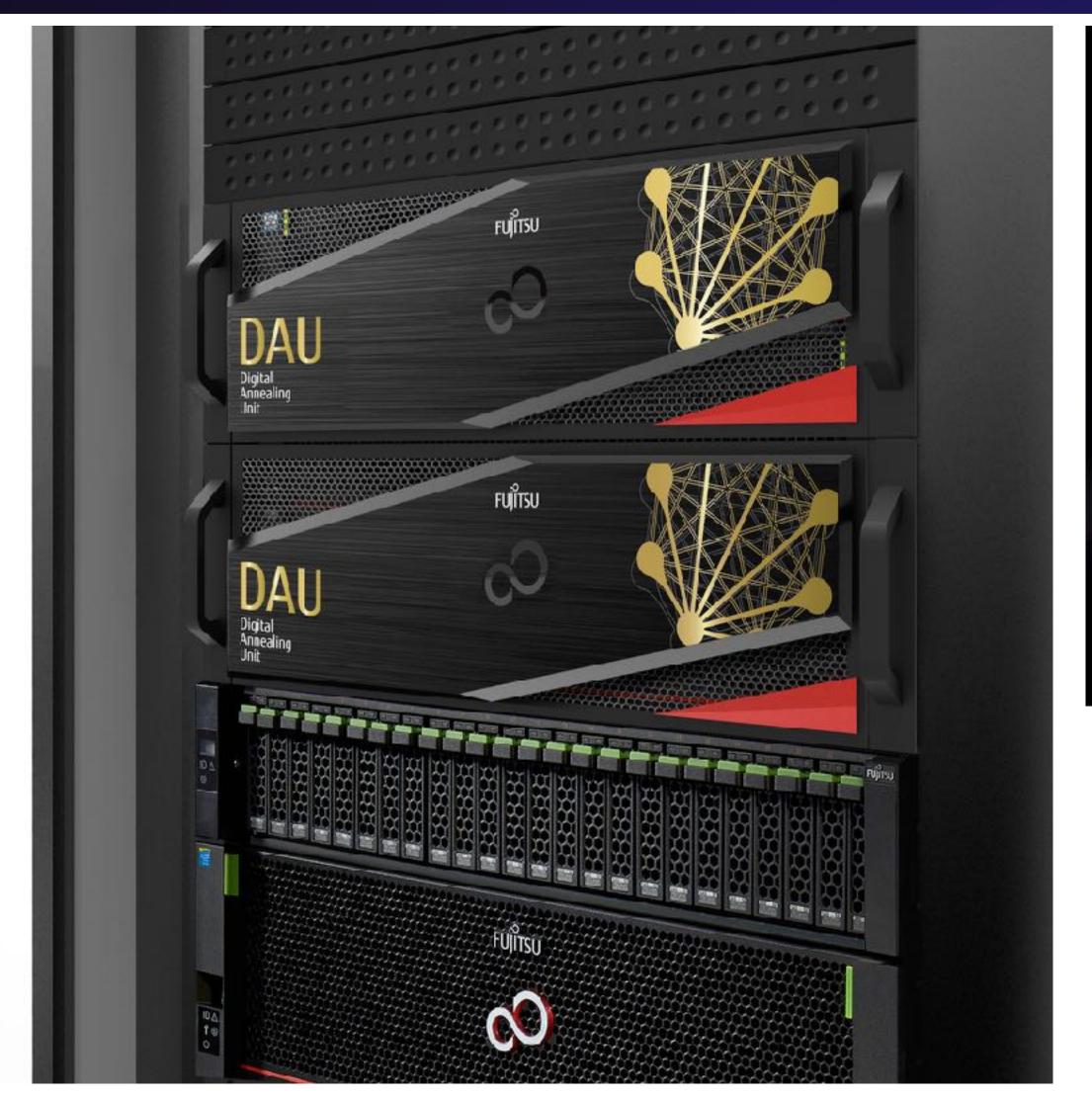


IBM True North



Intel Loihi

Digital Annealers

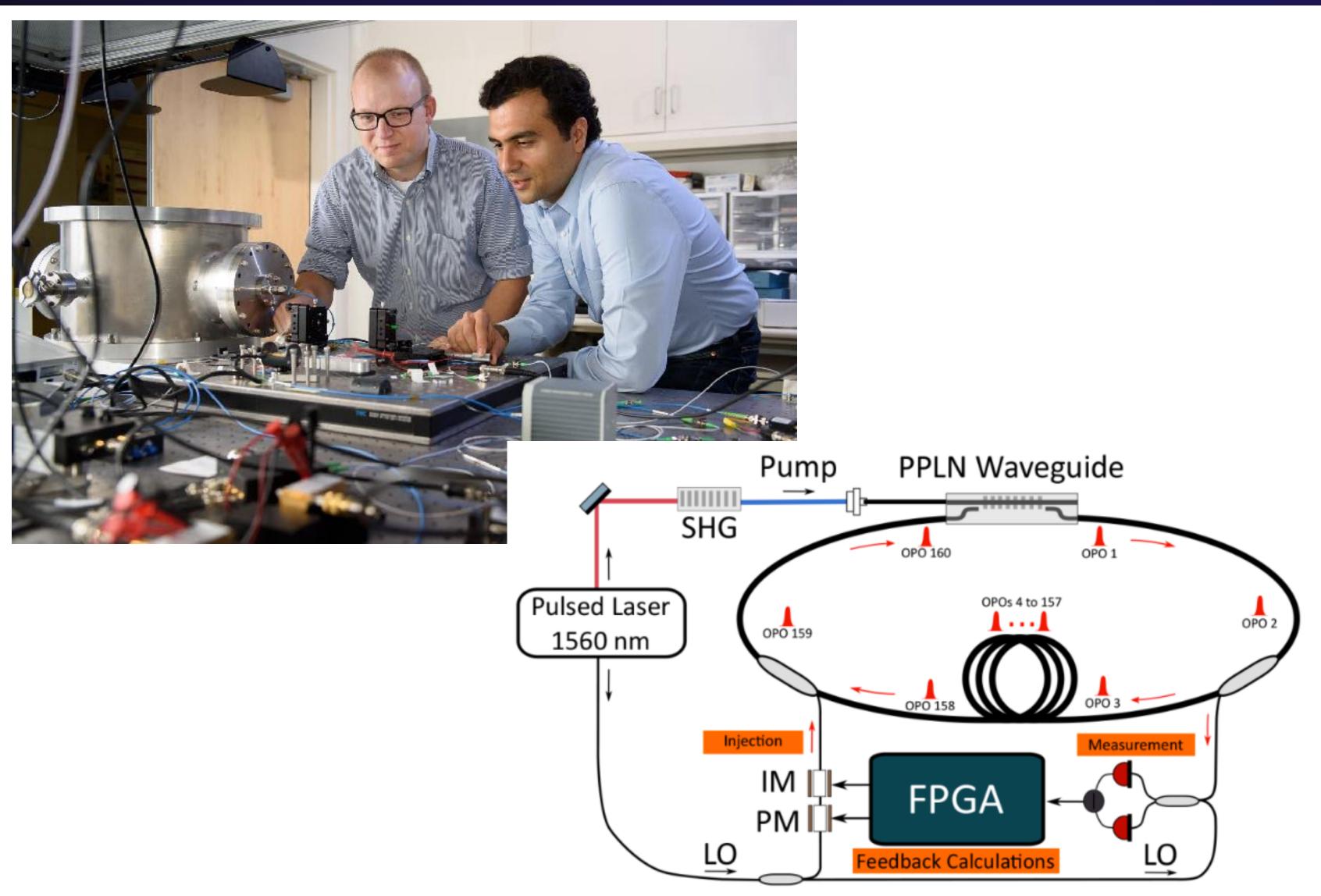


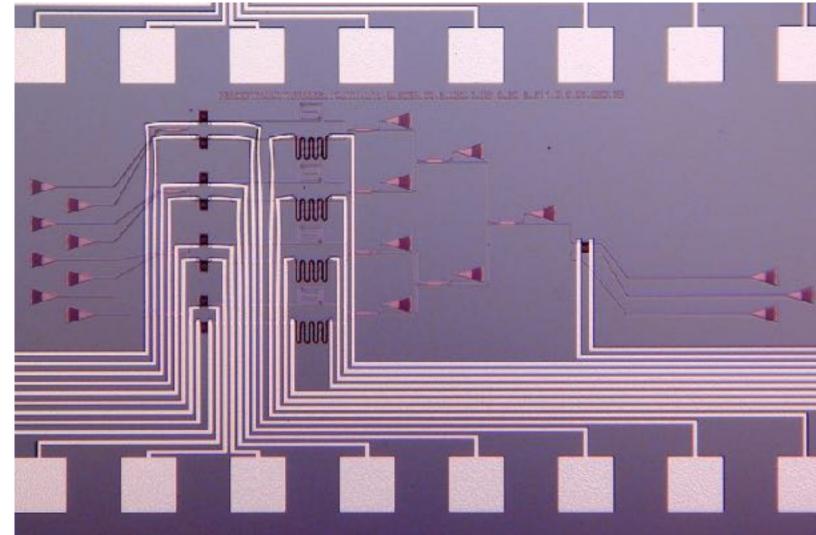




http://www.fujitsu.com/global/digitalannealer/

Optical Parametric Oscillators

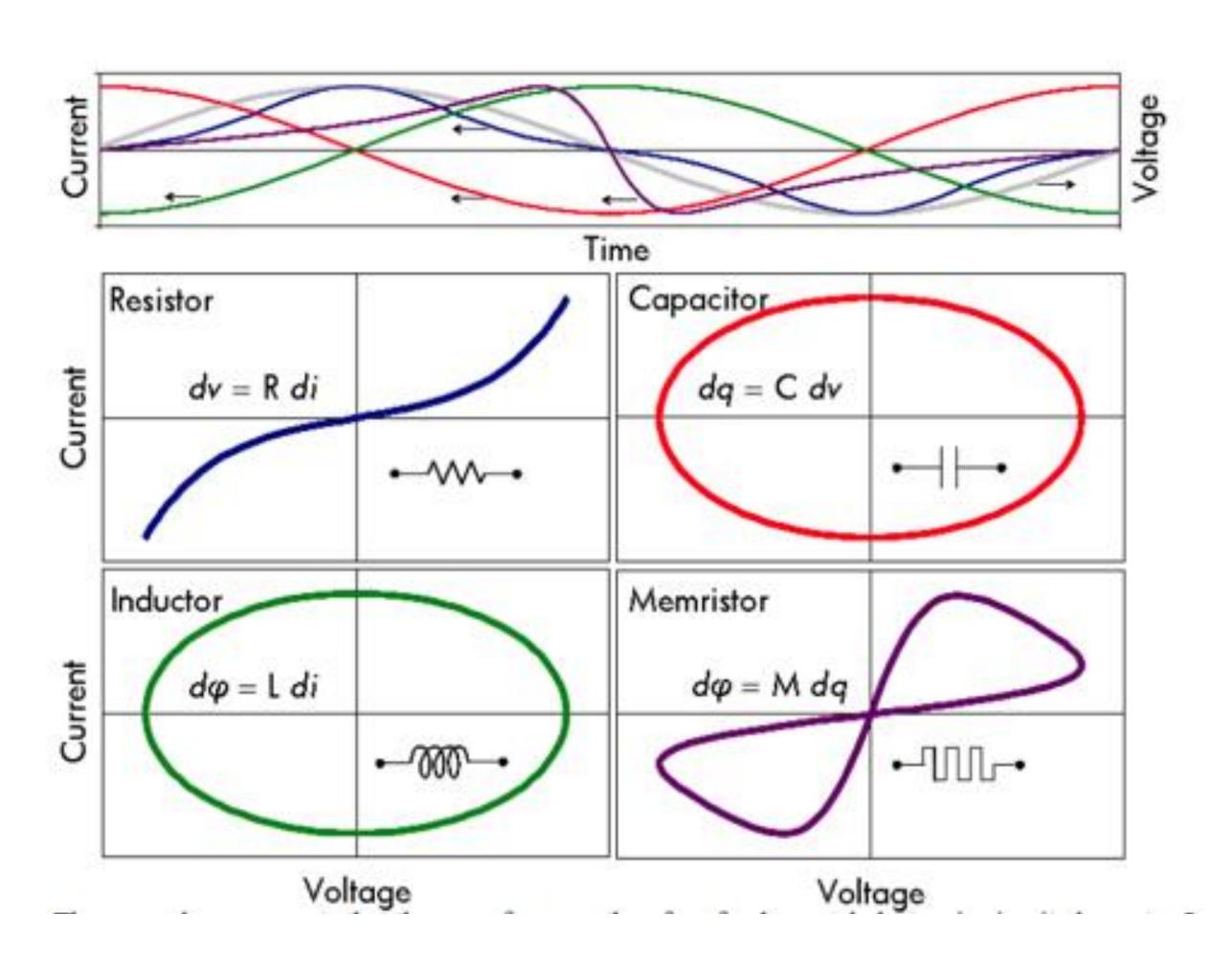




Close-up view of a complex optical circuit created by Hewlett Packard Labs.



Memristor Networks



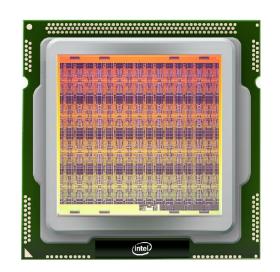




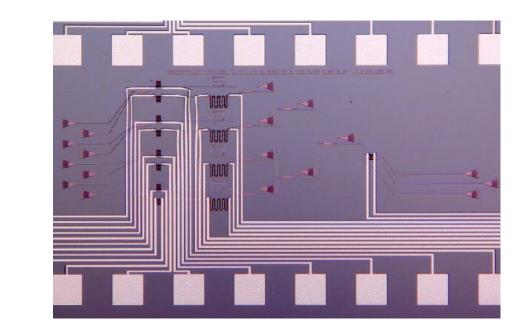




The Good News









QUBO

 $\min: \sum c_{ij}x_ix_j + \sum c_ix_i \quad \min: \sum c_{ij}\sigma_i\sigma_j + \sum c_i\sigma_i$ $i{\in}\mathcal{N}$ $i,j \in \mathcal{E}$

Ising

$$\sum oldsymbol{c}_{ij}\sigma_i\sigma_j + \sum oldsymbol{c}_i\sigma_i$$

 $i{\in}\mathcal{N}$



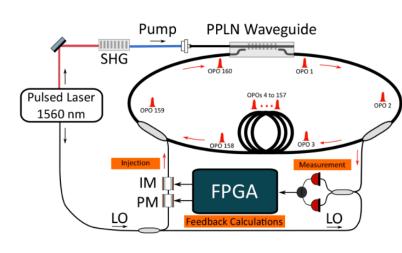
s.t.

$$x_i \in \{0, 1\} \ \forall i \in \mathcal{N}$$

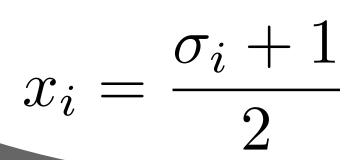
s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

 $i,j{\in}\mathcal{E}$

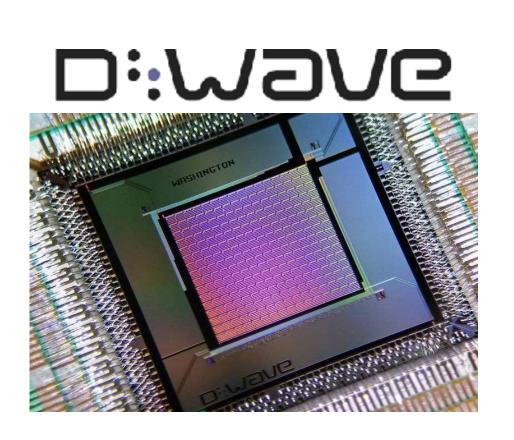






Ising Processing Units: Potential and Challenges for Discrete Optimization

https://arxiv.org/pdf/1707.00355.pdf

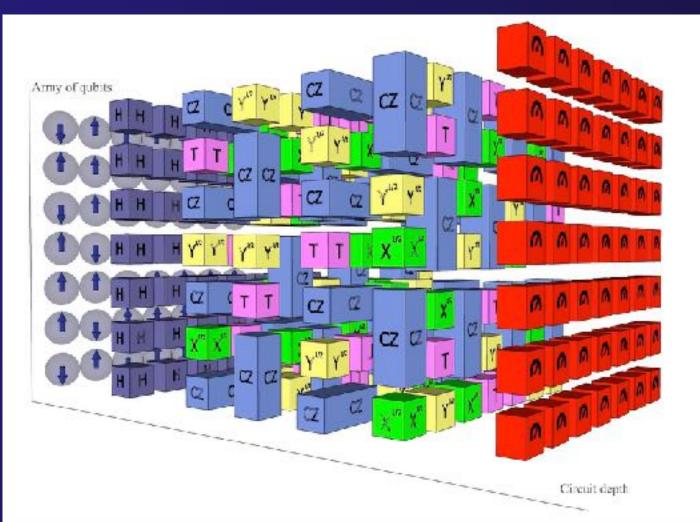


The Ambition of Novel Computing

Computational Supremacy

"[Can a novel computing device] perform a well-defined computational task beyond the capabilities of state-of-the-art classical computers" Boixo et. al.





NOTE: Initial computational tasks are likely to be very contrived!

The State of Novel Computing



UNIVAC 1960

The State of Novel Computing

What does it compute?

How fast does it compute? (esp. compared to state-of-the-art)

Overview

- My experience trying to make sense of a D-Wave Quantum Annealer
- What does it compute?
- Benchmarking Successes and Failures
- Future Outlooks

Ising

$$\min: \sum_{i,j \in \mathcal{E}} c_{ij} \sigma_i \sigma_j + \sum_{i \in \mathcal{N}} c_i \sigma_i$$

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$



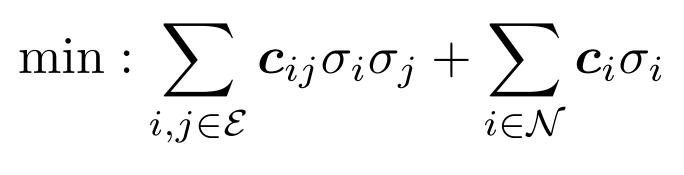
What does a D-Wave Compute?

The User Perspective

Ising Model Specification

Runtime Parameters (e.g. # of replicates)

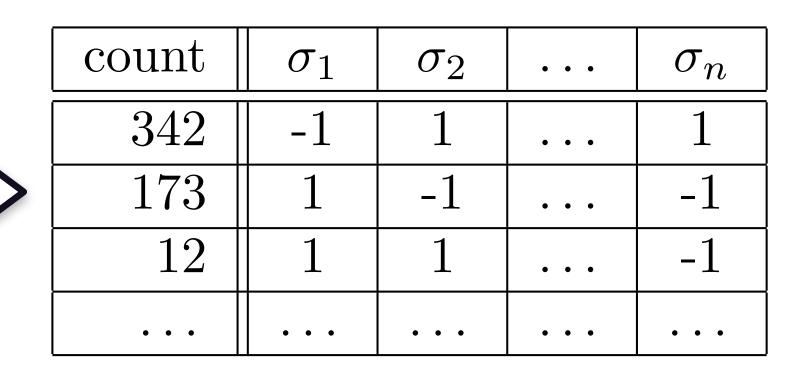
Variable Assignments



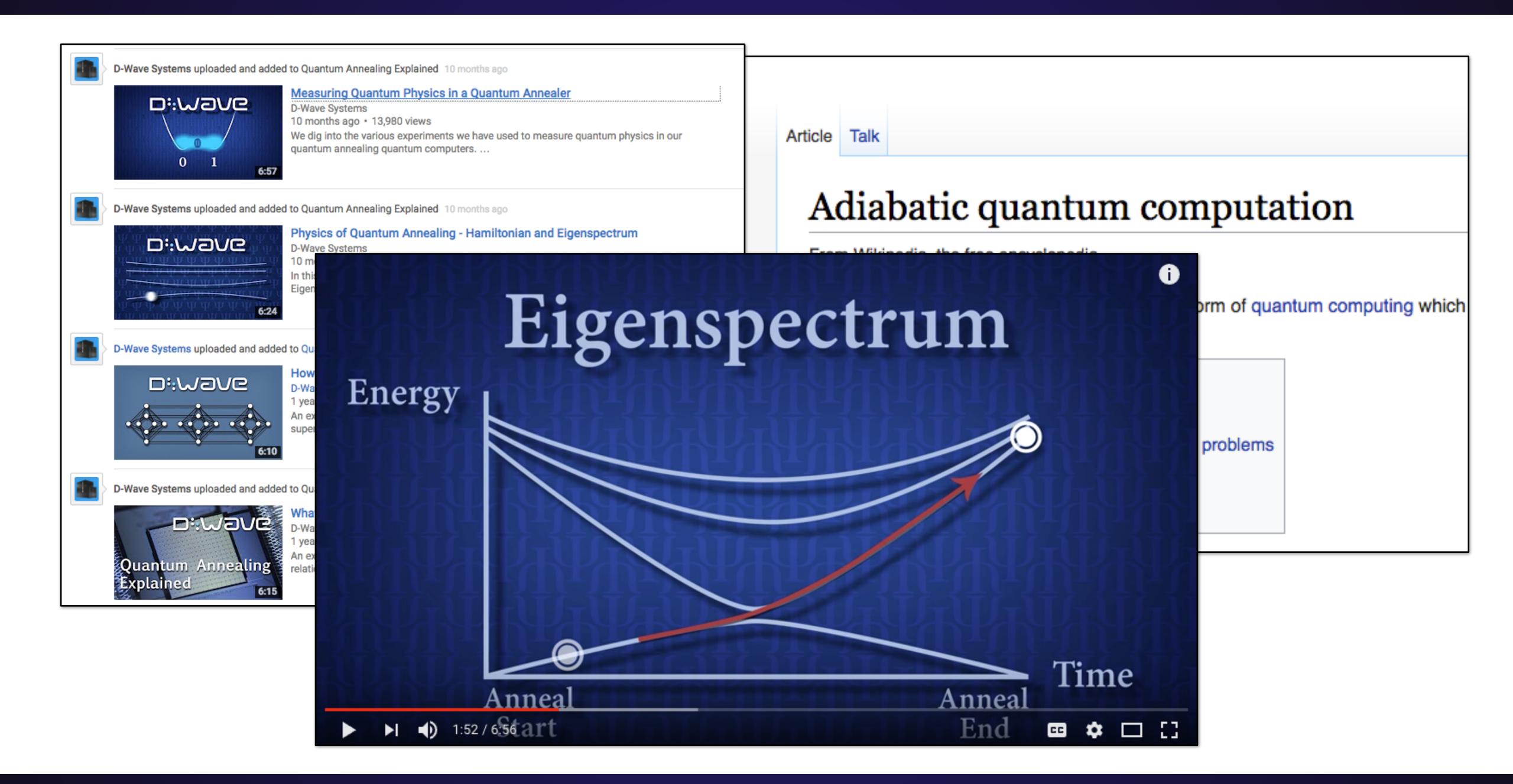
s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$





Ground Truth...



First Order Approximation

Finds globally optimal solutions to the Ising Model,

So what?

$$\min: \sum_{i,j \in \mathcal{E}} c_{ij}\sigma_i\sigma_j + \sum_{i \in \mathcal{N}} c_i\sigma_i$$

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

Max-Cut

$$\min: \sum_{i,j \in \mathcal{E}} \frac{\sigma_i \sigma_j - 1}{2}$$

WOW!

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

Let's Go!

Biq Mac Library - Binary quadratic and Max cut Library

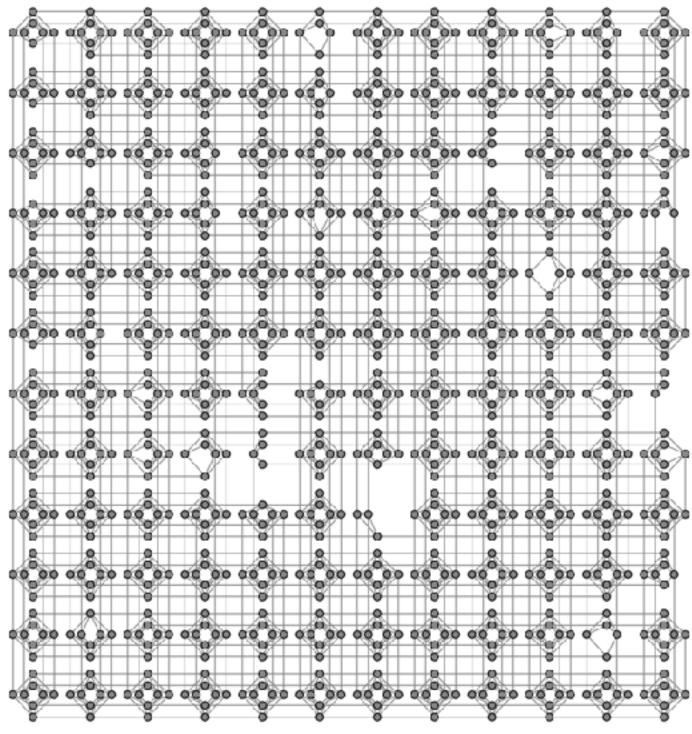
This site offers a collection of Max-Cut instances and quadratic 0-1 programming problems of medium size. Most of the instances were collected while developing <u>Biq Mac</u>, an SDP based Branch & Bound code (see [RRW07] or [Wie06]). The dimension of the problems (i.e., number of variables or number of vertices in the graph) ranges from 20 to 500. The instances are mainly ment to be used for testing exact solution methods for quadratic 0-1 programming or Max-Cut problems.

Any comments or further instances to be added are welcome! Please contact <u>angelika.wiegele@aau.at</u>.

The structure of the directories is as follows:

Some challenges...

Catch #1: Hardware Graph



 \mathcal{C}_{12}

DW2X

$$\min: \sum_{i,j \in \mathcal{E}} c_{ij}\sigma_i\sigma_j + \sum_{i \in \mathcal{N}} c_i\sigma_i$$

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

$$\mathcal{E} \subseteq \mathcal{C}_{12}$$

$$|\mathcal{N}| \le 1100$$

$$|\mathcal{E}| \le 3068$$

Catch #2: Coefficient Values

$$c_{ij} \in \{-1.00, -0.99, -0.98, \dots, 0.98, 0.99, 1.00\} \ \forall i, j \in \mathcal{E}$$

 $c_i \in \{-2.00, -1.99, -1.98, \dots, 1.98, 1.99, 2.00\} \ \forall i \in \mathcal{N}$

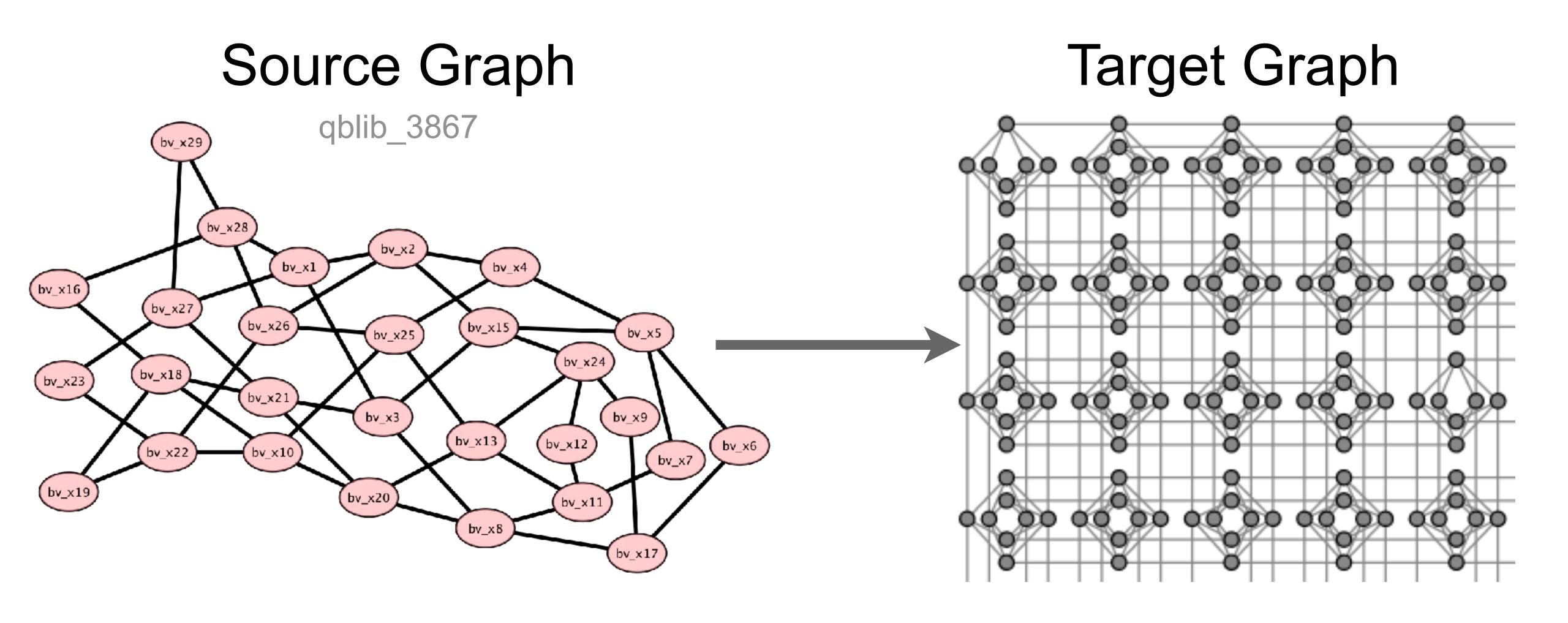
$$\min: \sum_{i,j \in \mathcal{E}} c_{ij}\sigma_i\sigma_j + \sum_{i \in \mathcal{N}} c_i\sigma_i$$

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

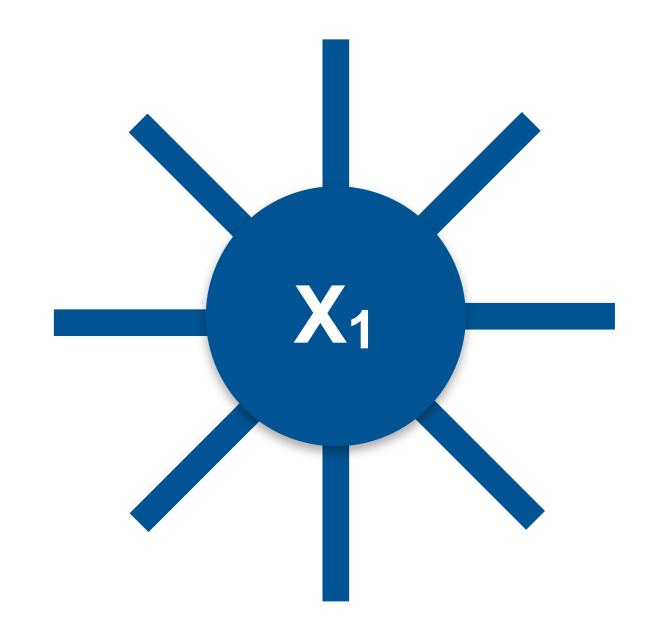
$$\mathcal{E} \subseteq \mathcal{C}_{12}$$

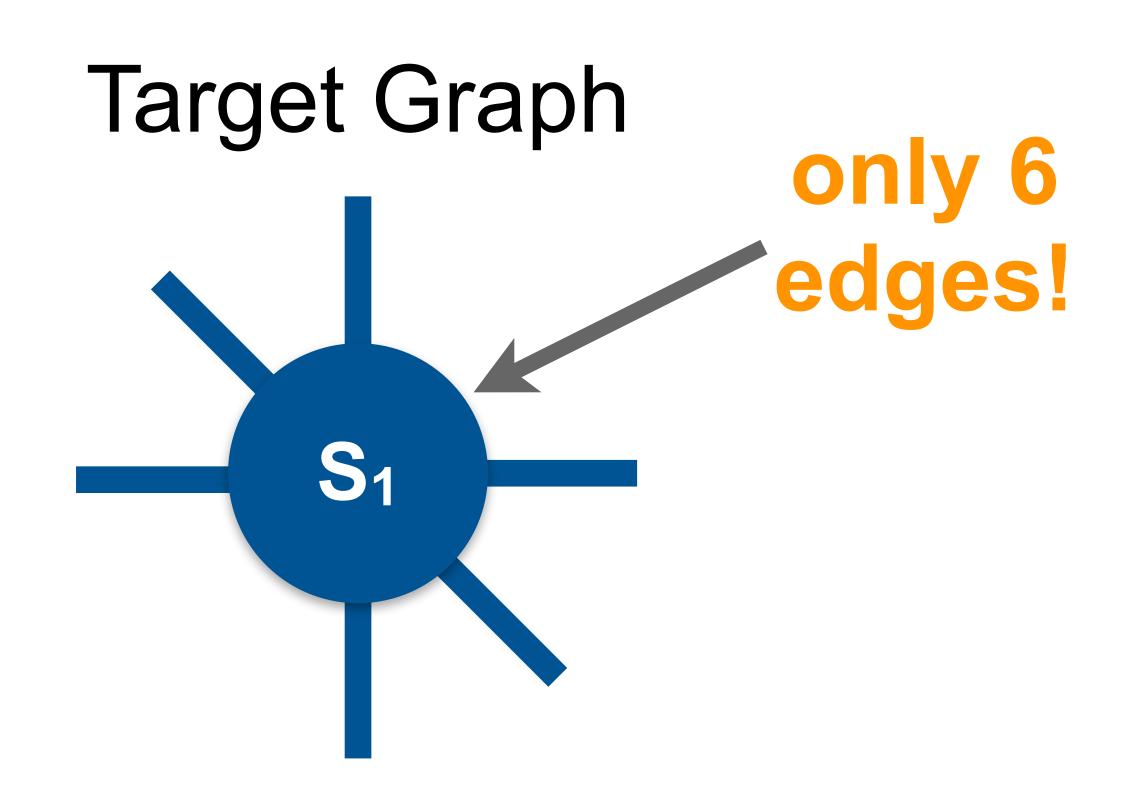
Catch #3: Graph Embedding



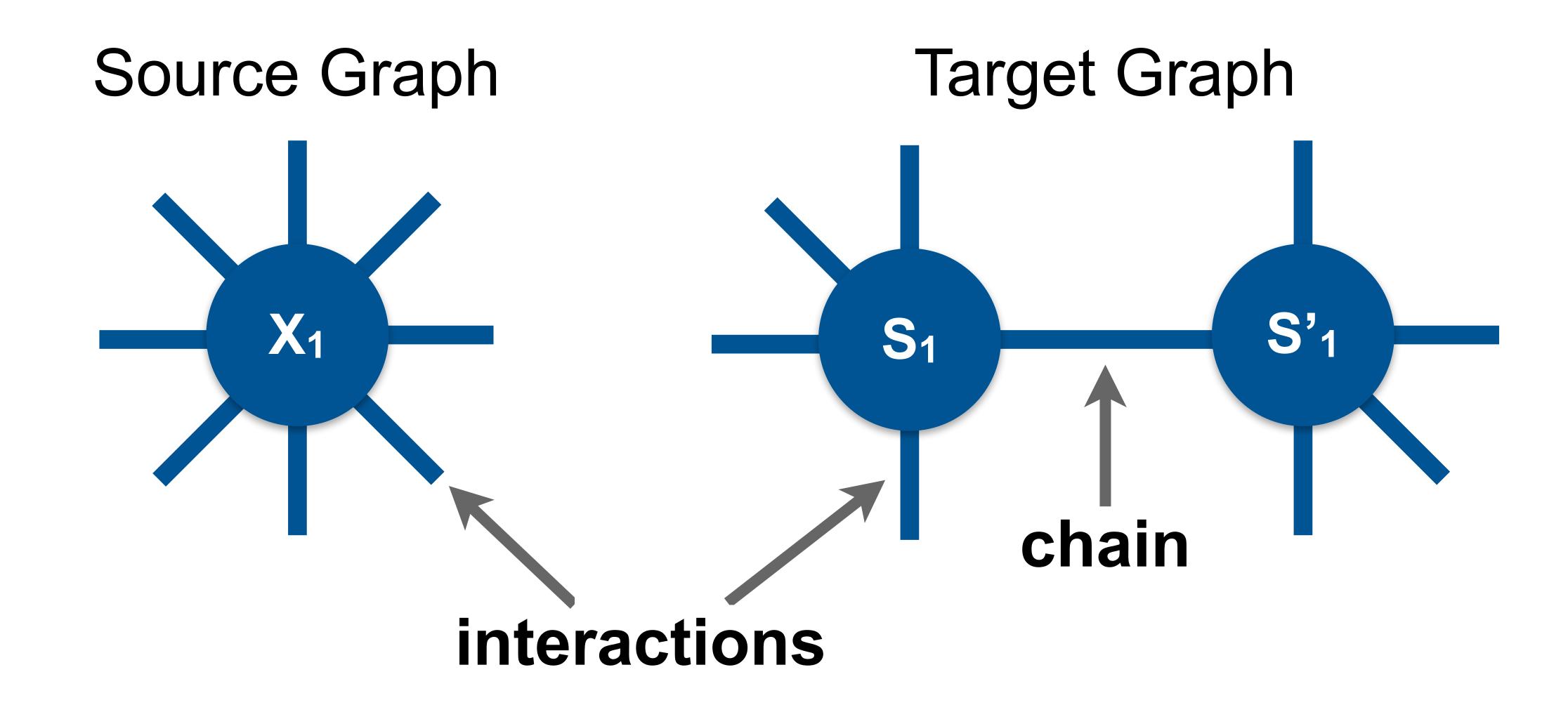
Catch #3: Graph Embedding

Source Graph



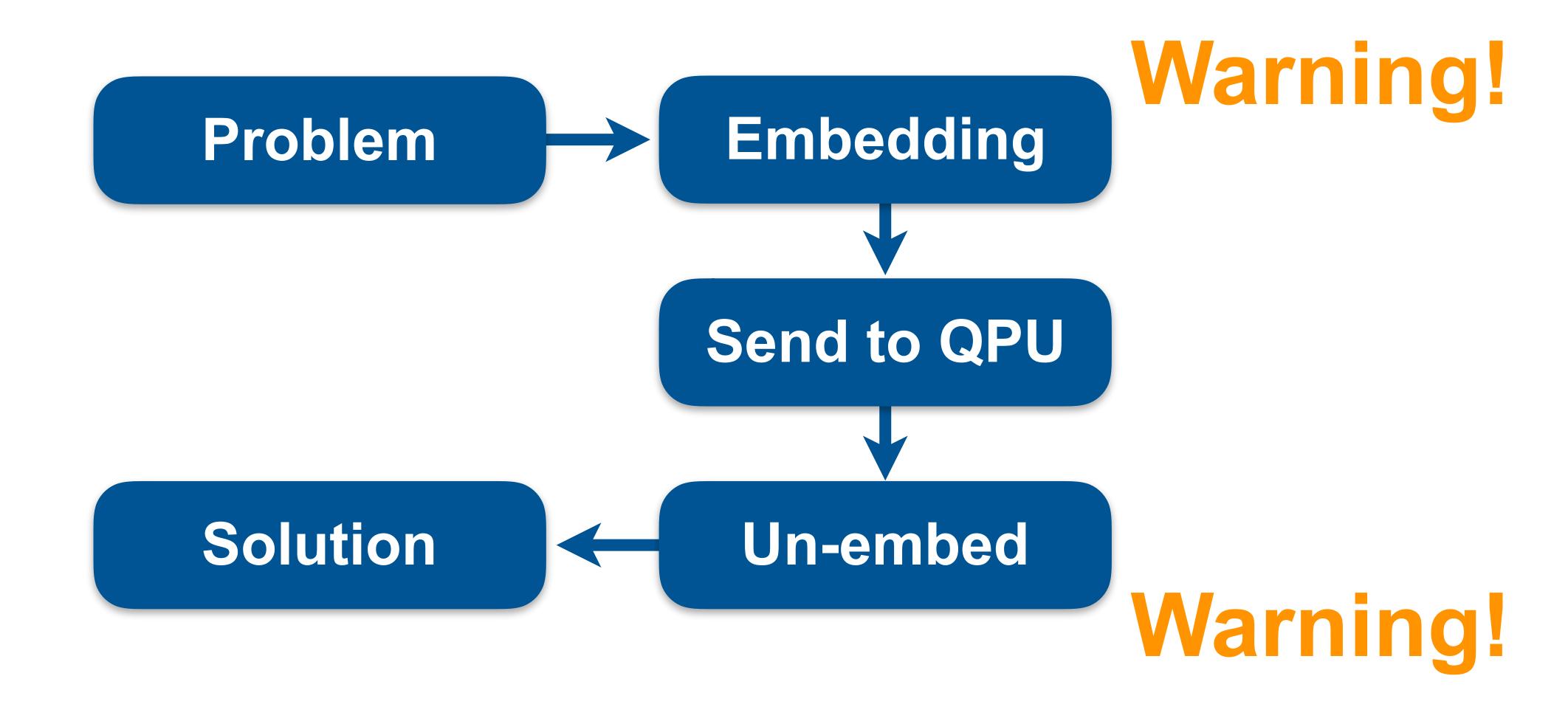


Catch #3: Graph Embedding

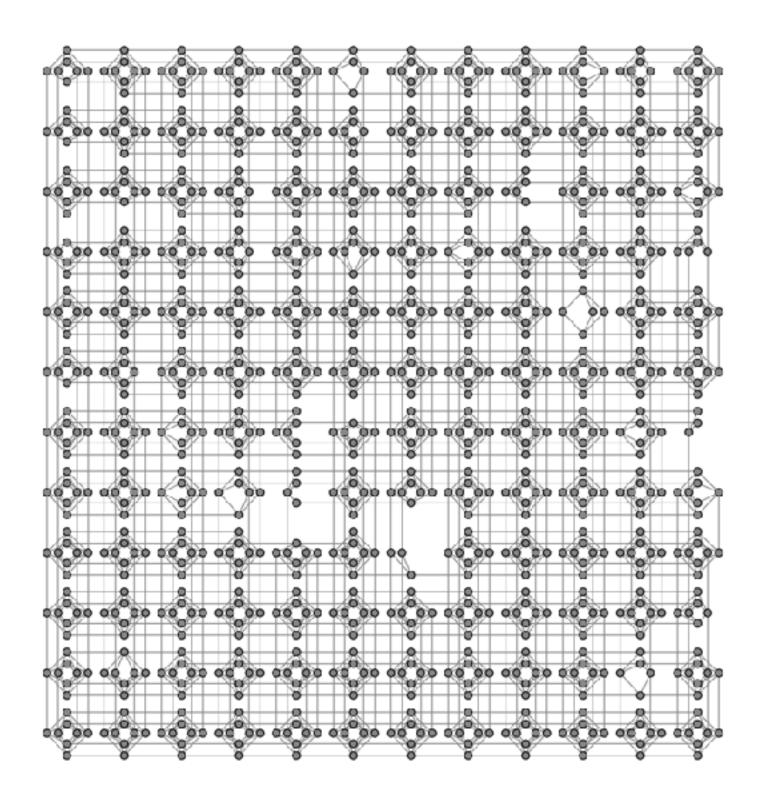


Warning: broken chains = infeasible solution

Typical D-Wave Algorithm

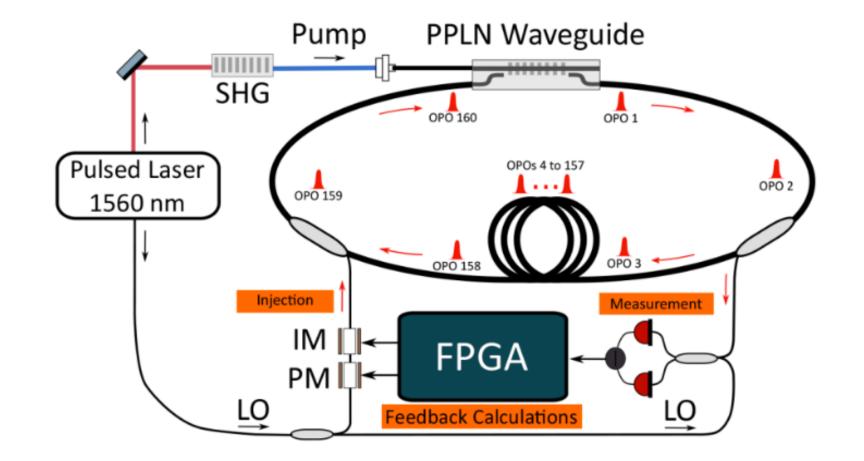


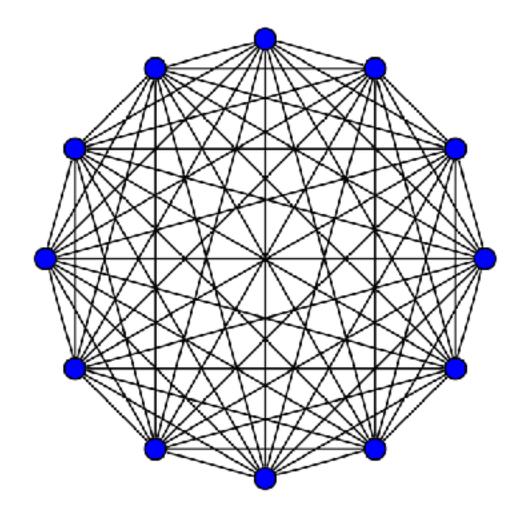
A Note on Other Technologies



$$c_{ij} \in \{-1.00, -0.99, -0.98, \dots, 0.98, 0.99, 1.00\} \ \forall i, j \in \mathcal{E}$$

 $c_i \in \{-2.00, -1.99, -1.98, \dots, 1.98, 1.99, 2.00\} \ \forall i \in \mathcal{N}$





$$c_{ij} \in \{-1, 0, 1\} \ \forall i, j \in \mathcal{E}$$

$$c_i \in \{-1, 0, 1\} \ \forall i \in \mathcal{N}$$

Back to Benchmarking

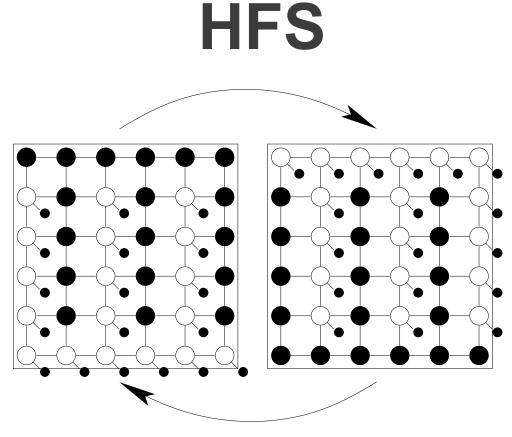
Classic Operations Research Approach

- Choose a widely recognized benchmark library
 - QPLib
 - DIMACS Max-Clique Cases



- Measure runtime against state-of-the-art alternatives
 - Complete Search (e.g. MIP)
 - Local Search Heuristics (e.g. HFS)





MIP-Based Ising Model Solver

Ising Model

QUBO Model

ILP Model

$$\min: \sum_{i,j \in \mathcal{E}} c_{ij} \sigma_i \sigma_j + \sum_{i \in \mathcal{N}} c_i \sigma_i$$

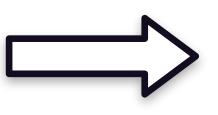
$$\min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij}\sigma_i\sigma_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_i\sigma_i \quad \min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij}x_ix_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_ix_i \quad \min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij}x_{ij} + \sum_{i\in\mathcal{N}} \boldsymbol{c}_ix_i$$

$$\min: \sum_{i,j\in\mathcal{E}} c_{ij} x_{ij} + \sum_{i\in\mathcal{N}} c_i x_i$$

s.t.

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N} \quad \square \qquad x_i \in \{0, 1\} \ \forall i \in \mathcal{N}$$

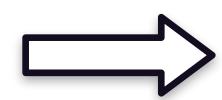
$$\mathcal{E} \subseteq \mathcal{C}_{12}$$



$$x_i \in \{0, 1\} \ \forall i \in \Lambda$$

$$\mathcal{E} \subseteq \mathcal{C}_{12}$$

s.t.



$$2x_{ij} \geq x_i + x_j \ \forall i, j \in \mathcal{E}$$

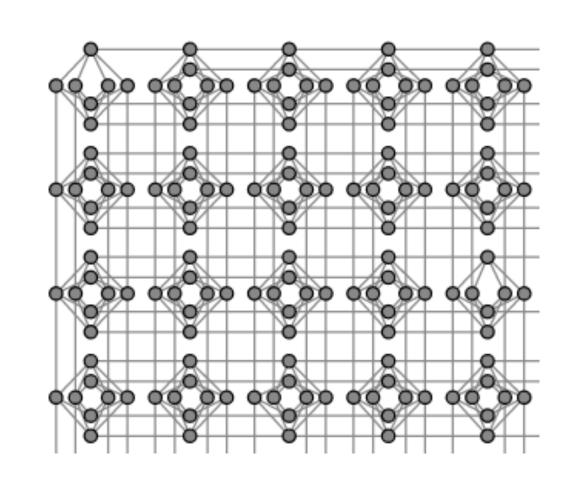
$$x_{ij} \leq x_i \ \forall i, j \in \mathcal{E}$$

$$x_{ij} \leq x_j \ \forall i, j \in \mathcal{E}$$

$$x_{ij} \in \{0,1\} \ \forall i,j \in \mathcal{E}$$

$$x_i \in \{0, 1\} \ \forall i \in \mathcal{N}$$

$$\mathcal{E} \subseteq \mathcal{C}_{12}$$





Billionnet, Alain, and Sourour Elloumi. "Using a mixed integer quadratic programming solver for the unconstrained quadratic 0-1 problem." Mathematical Programming 109.1 (2007): 55-68.

HFS Local Search Solver

Hamze-de Freitas-Selby (HFS)

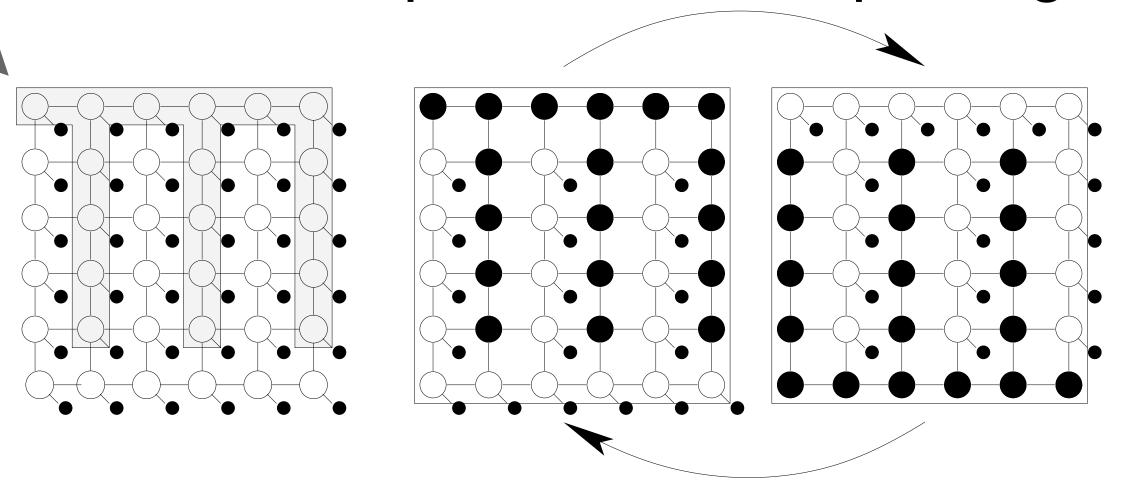
https://arxiv.org/pdf/1207.4149.pdf

https://github.com/alex1770/QUBO-Chimera

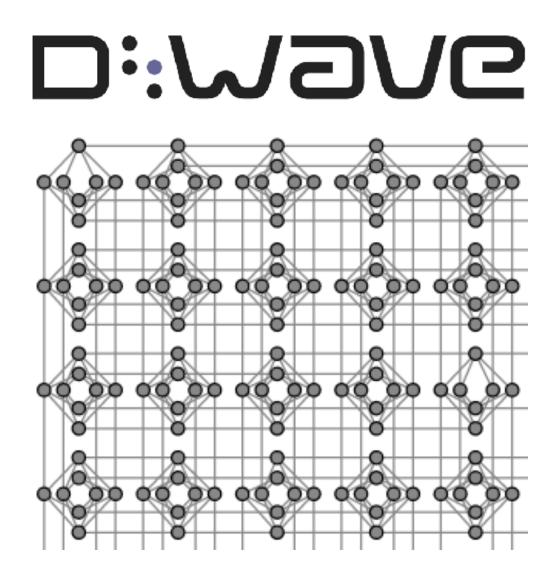
Low Treewidth

Subgraphs

Optimization Loop using DP



highly optimized C code



HF

S

Benchmark #1: QPLIB

MI-QCQP

$$\min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij} x_i x_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_i x_i$$

s.t.

$$\sum_{i,j\in\mathcal{E}} c_{sij} x_i x_j + \sum_{i\in\mathcal{N}} c_{si} x_i \le 0 \ \forall s \in \mathcal{S}$$

$$x_i \in \{0, 1\} \ \forall i \in \mathcal{B} \subseteq \mathcal{N}$$

$$x_i \in \mathbb{N} \ \forall i \in \mathcal{I} \subseteq \mathcal{N}$$

$$x_i \in \mathbb{R} \ \forall i \in \mathcal{R} \subseteq \mathcal{N}$$



QUBO Model

$$\min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij} x_i x_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_i x_i$$

s.t.

$$x_i \in \{0, 1\} \ \forall i \in \mathcal{N}$$

$$|\mathcal{N}| \le 1100$$

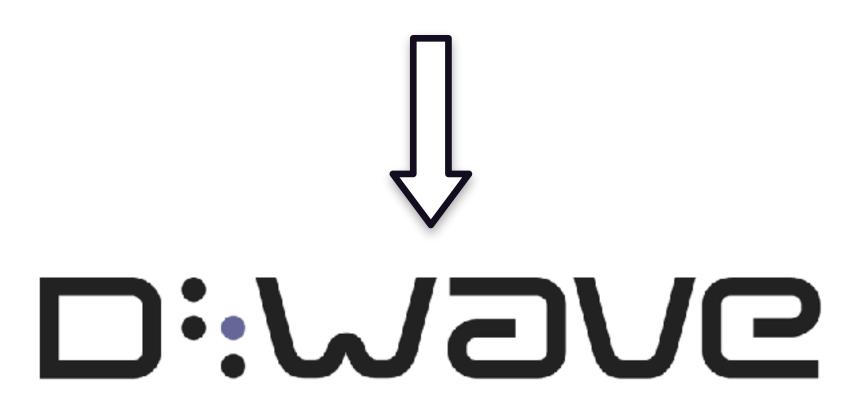
$$|\mathcal{E}| \leq 3068$$



Benchmark #1: QPLIB

- 479 cases
 - -350 cases, no real or int
 - -87 cases, no constraints
 - -23 cases, edge bound
 - -2 cases, variable bound
 - 17 cases remain...





Benchmark #1: QPLIB

	gurobi				dwave						
Case	$ \mathcal{N} $	$ \mathcal{E} $	Best Sol.	Opt	G. Gap	Time	Best Sol.	Best	Inf.	Samples	Time
qplib_3876	28	48	-24		0%	<1	-24	8574	352	10000	0+4
qplib_3607	66	120	-68		0%	<1	-68	1720	4424	10000	1+4
qplib_5727	225	450	-15051133		0%	<1	F.E.	_	_	_	T.L.
qplib_3756	153	288	-160		0%	1.2	-160	234	8451	10000	57+5
qplib_5755	400	800	-24838942		0%	1.5	F.E.	_	-	-	T.L.
qplib_3821	190	360	-192		0%	15	F.E.	-	-	-	T.L.
qplib_3565	276	528	-282		0%	93	F.E.	-	-	_	T.L.
qplib_3705	378	728	-384		0%	202	F.E.	-	-	_	T.L.
qplib_3745	325	624	-334		0%	753	F.E.	-	-	_	T.L.
qplib_3506	496	960	-478		1%	T.L.	F.E.	_	-	_	T.L.
qplib_3738	435	840	-422		1%	T.L.	F.E.	_	-	-	T.L.
qplib_3877	630	1224	-602		1%	T.L.	F.E.	_	-	_	T.L.
qplib_3706	703	1368	-682		2%	T.L.	F.E.	_	-	_	T.L.
qplib_3642	1035	2024	-1030		3%	T.L.	F.E.	_	-	_	T.L.
qplib_3650	946	1848	-918		3%	T.L.	F.E.	-	-	_	T.L.
qplib_5889	250	3045	-40358		15%	T.L.	F.E.*	_	_	-	T.L.
qplib_5909	250	3015	-33587		28%	T.L.	F.E.*	_	_	_	T.L.

F.E.
Failed Embed
T.L.
Time Limit

(1 hour)

Benchmark #2: DIMACS Max-Clique

http://iridia.ulb.ac.be/~fmascia/maximum_clique/DIMACS-benchmark

dimacs benchmark set

This is a selected set of instances form the Second DIMACS Implementation Challenge (1992-1993).

In the following table, the clique number $\omega(G)$ corresponds to the global optimum or to the lower bound as indicated by the For some instances the lower bound has been confirmed to coincide with the global optimum. For such instances, the clique The exact algorithms that confirmed the bound are reported later on this page.

instance	ω (g)	best known	nodes	odaoc	graph degrees		best degrees	
				edges	median	iqr	median	iqr
C125.9	34*	34	125	6 963	112.0	(5.00)	114.5	(4.75)
C250.9	44*	44	250	27 984	224.0	(6.00)	227.0	(5.00)
C500.9	≥ 57	57	500	112 332	449.0	(9.00)	455.0	(9.00)
C1000.9	≥68	68	1000	450 079	900.0	(13.00)	907.0	(11.25)
C2000.9	≥80	80	2 000	1 799 532	1 800.0	(18.00)	1 803.0	(15.25)
DSJC1000_5	15	15	1000	499 652	500.0	(20.00)	503.0	(23.00)
DSJC500_5	13	13	500	125 248	250.0	(16.00)	259.0	(14.00)
C2000.5	16*	16	2 000	999 836	999.0	(30.00)	1 006.0	(11.50)

Why Max-Clique?

- Formulate on the Compliment Graph
 - super sparse, i.e. easy to embed!
- Combinatorial Problem
 - minimal issues due to coefficient accuracy
- Easy to generate problems

Benchmark #2: DIMACS Max-Clique

				gurobi	dwave					
Case		$ \mathcal{E} $	Best Sol.	Opt. Gap	Time	Best Sol.	Best	_ Inf.	Samples	Time
C015_9	15	12	-11	0%	<1	-11	9073	1	10000	0+3
C020_9	20	17	-14	0%	<1	-14	8370	85	10000	0+3
C030_9	30	44	-16	0%	<1	-16	5651	123	10000	0+3
C040_9	40	77	-18	0%	<1	-18	3865	316	10000	0+4
C050_9	50	108	-24	0%	<1	-24	16	1254	10000	0+4
C060_9	60	158	-25	0%	<1	-25	22	5465	10000	0+5
C070_9	70	215	-27	0%	<1	-26	1	9855	10000	4+5
C080_9	80	306	-29	0%	<1	F.E.	-	-	_	T.L.
C090_9	90	407	-29	0%	1.0	F.E.	_	_	_	T.L.
C100_9	100	508	-30	0%	2.0	F.E.	_	_	_	T.L.
C110_9	110	615	-32	0%	5.1	F.E.	_	_	-	T.L.
C120_9	120	729	-32	0%	45	F.E.	-	_	-	T.L.
C125_9	125	787	-34	0%	55	F.E.		_	_	T.L.
$\boxed{\text{C250_9}}$	250	3141	-43	40%	T.L.	F.E.*	_	_	_	T.L.

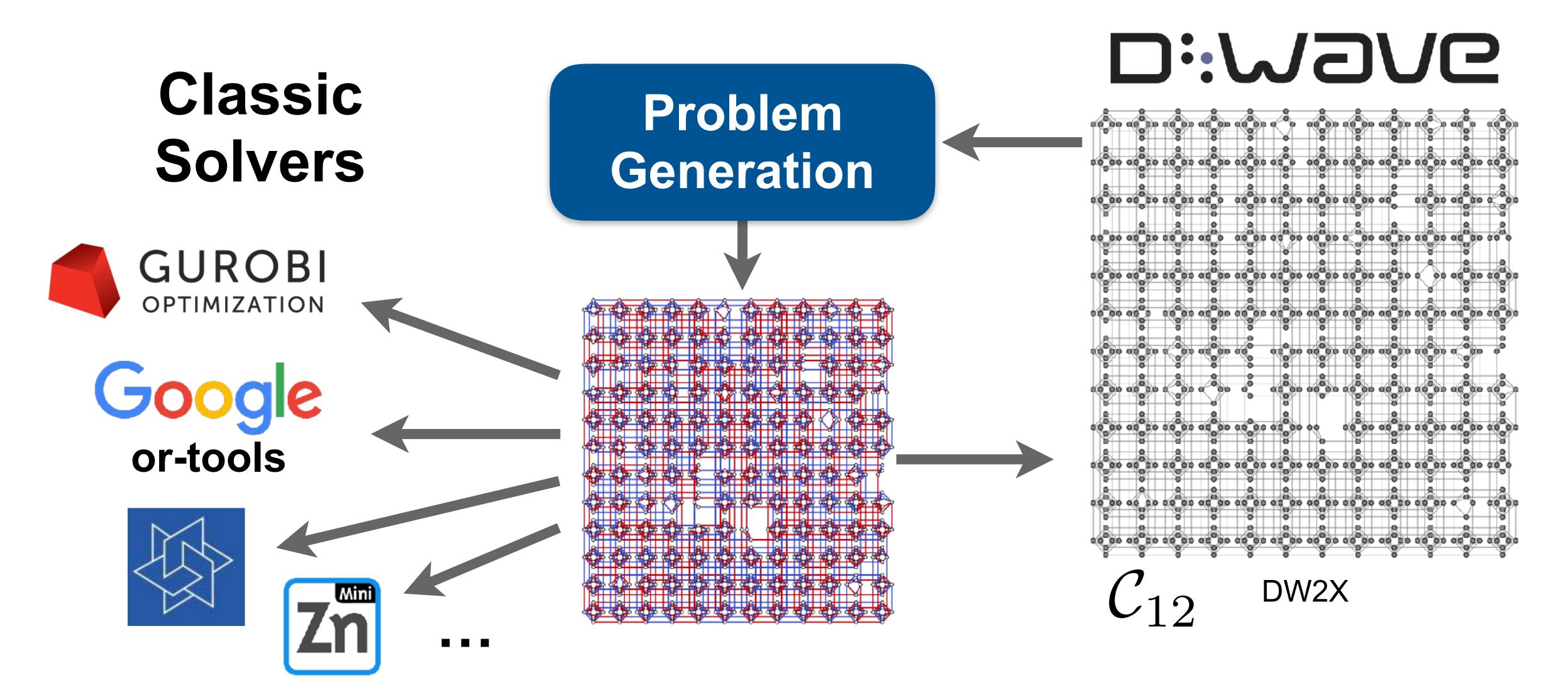
F.E.
Failed
Embed

T.L.
Time
Limit
(1 hour)

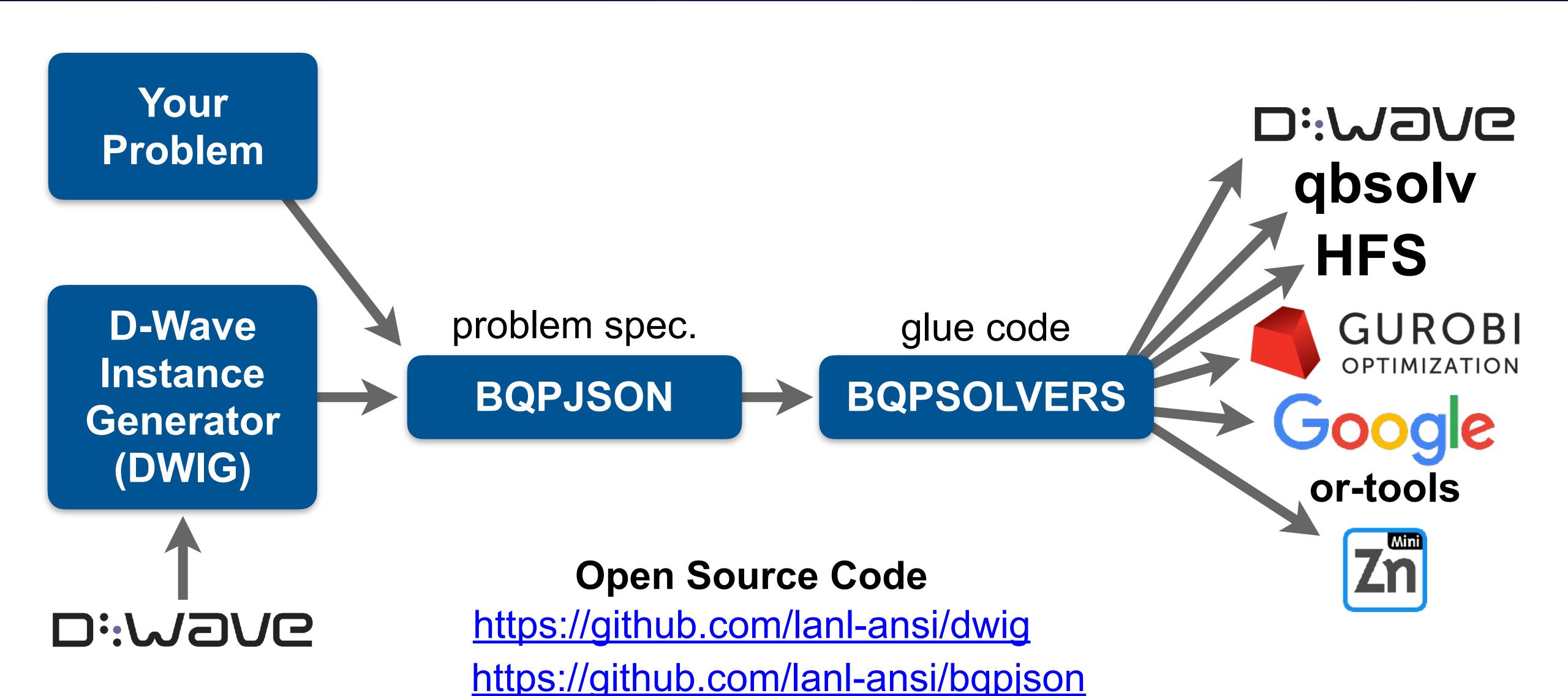
Back to the Drawing Board

Benchmarking Stopgap

• Embedding real problems is just too difficult, at the moment.



Benchmarking Tools



https://github.com/lanl-ansi/bqpsolvers

The Problem with Problem Generation

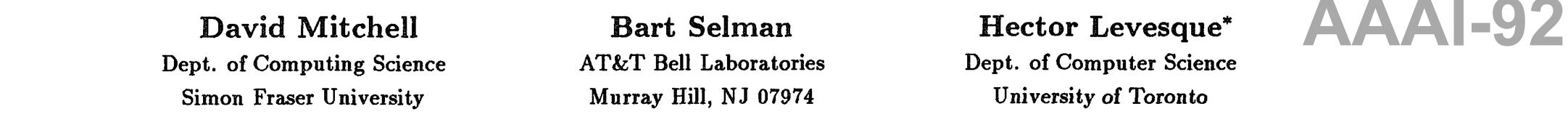
Generating interesting problems can be very hard!

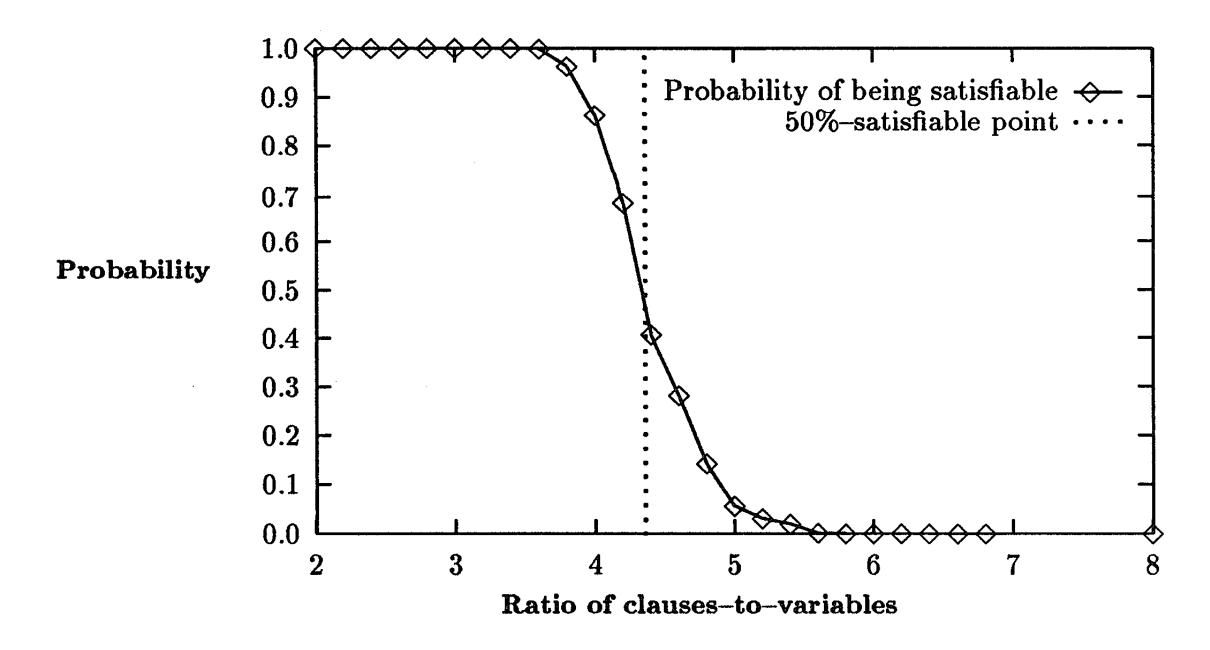
Hard and Easy Distributions of SAT Problems

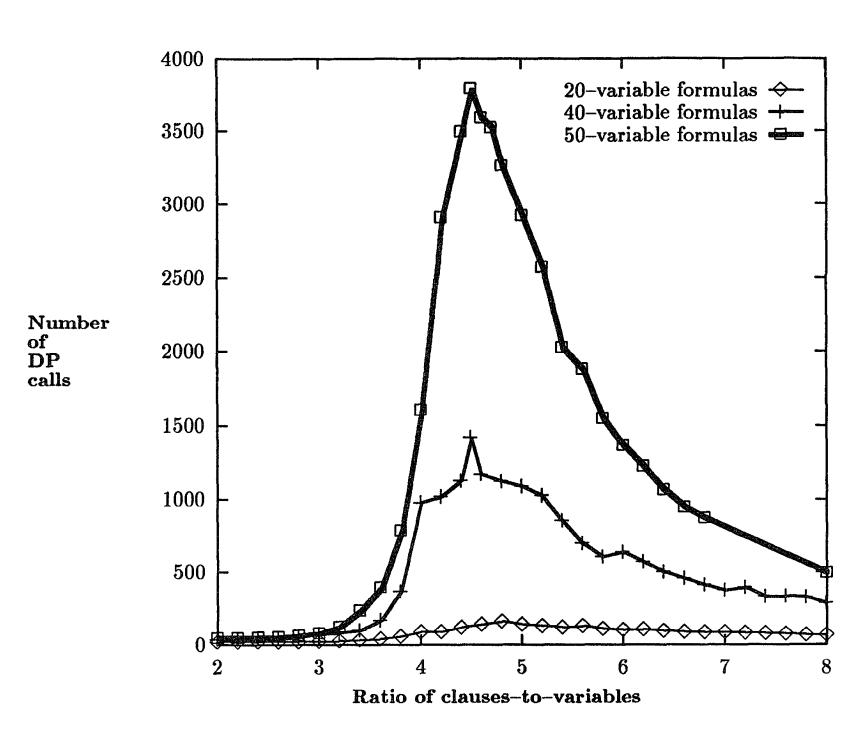
David Mitchell

Bart Selman

Hector Levesque*

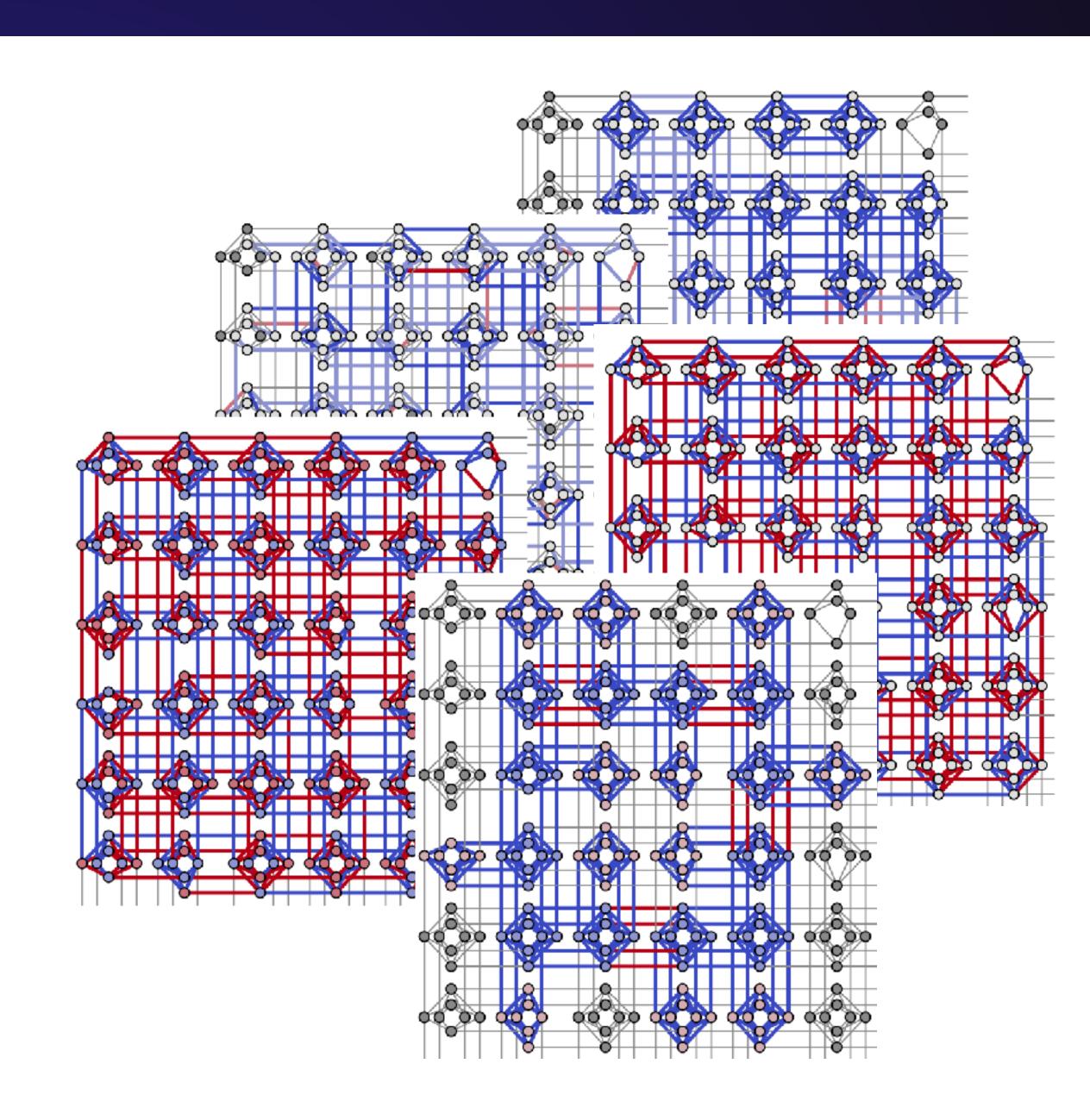




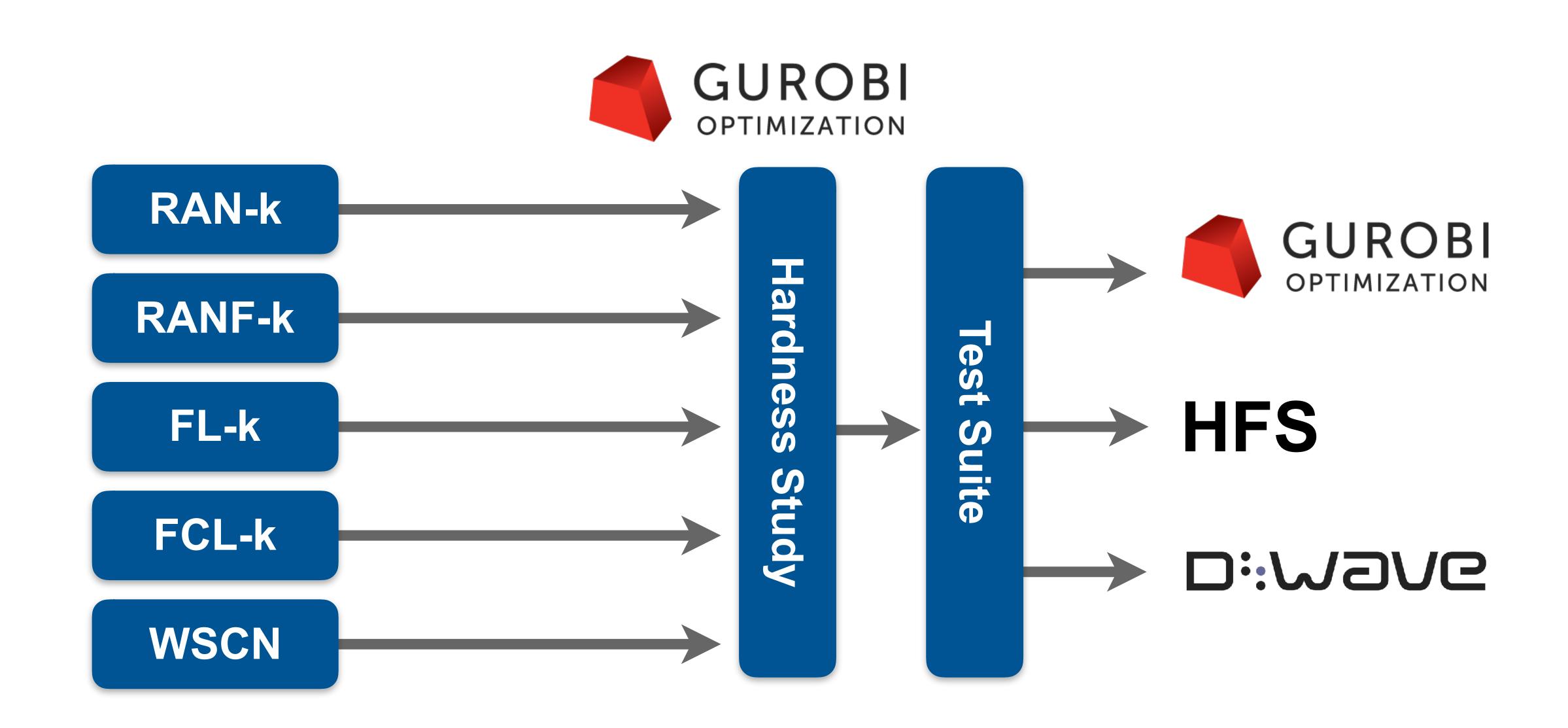


What Problems to Consider

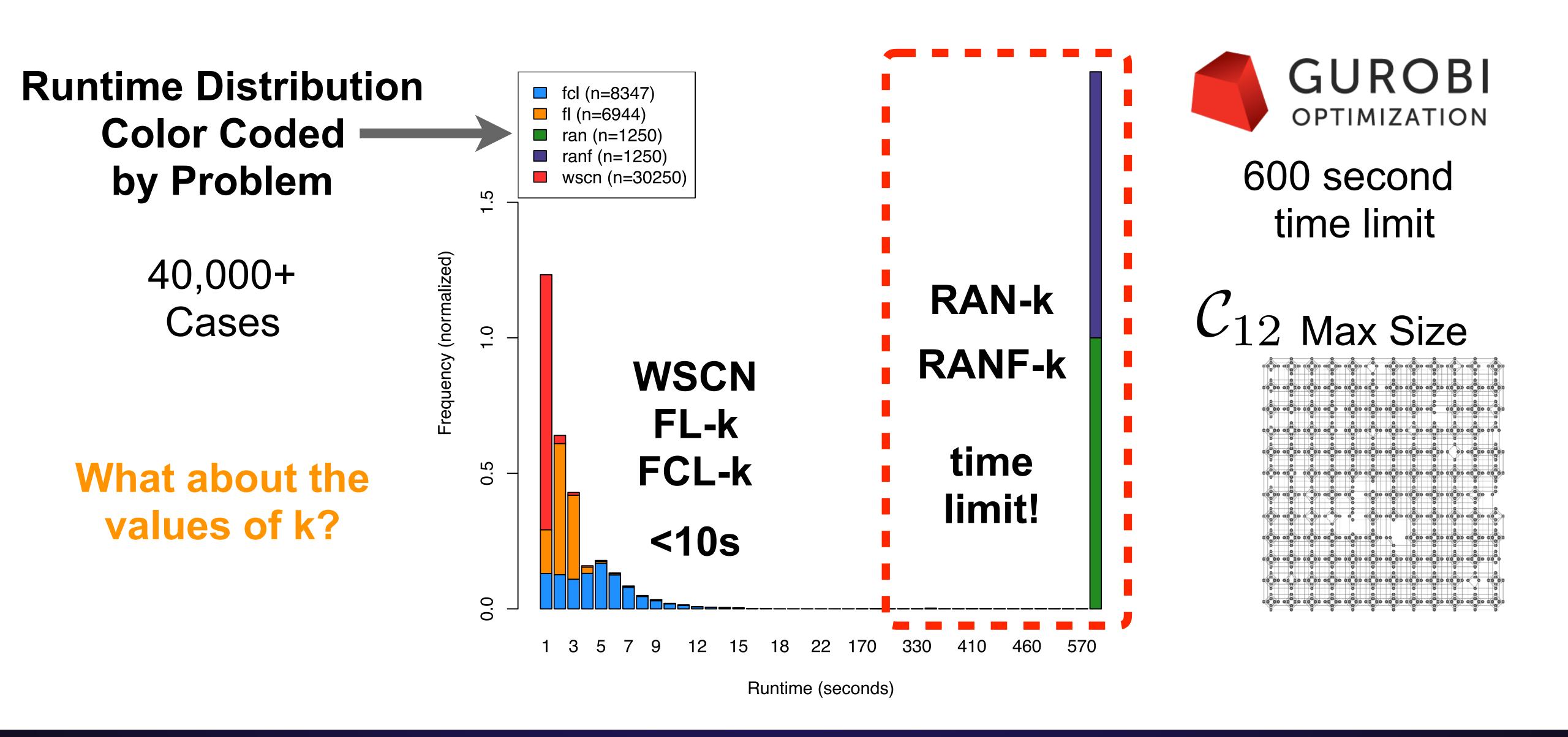
- Try everything from the literature
- RAN-k, RANF-k
 - https://arxiv.org/abs/1508.05087
- Frustrated Loops
 - https://arxiv.org/abs/1508.05087
 - https://arxiv.org/abs/1701.04579
 - https://arxiv.org/abs/1703.00622
- Weak-Strong Cluster Networks
 - https://arxiv.org/abs/1512.02206



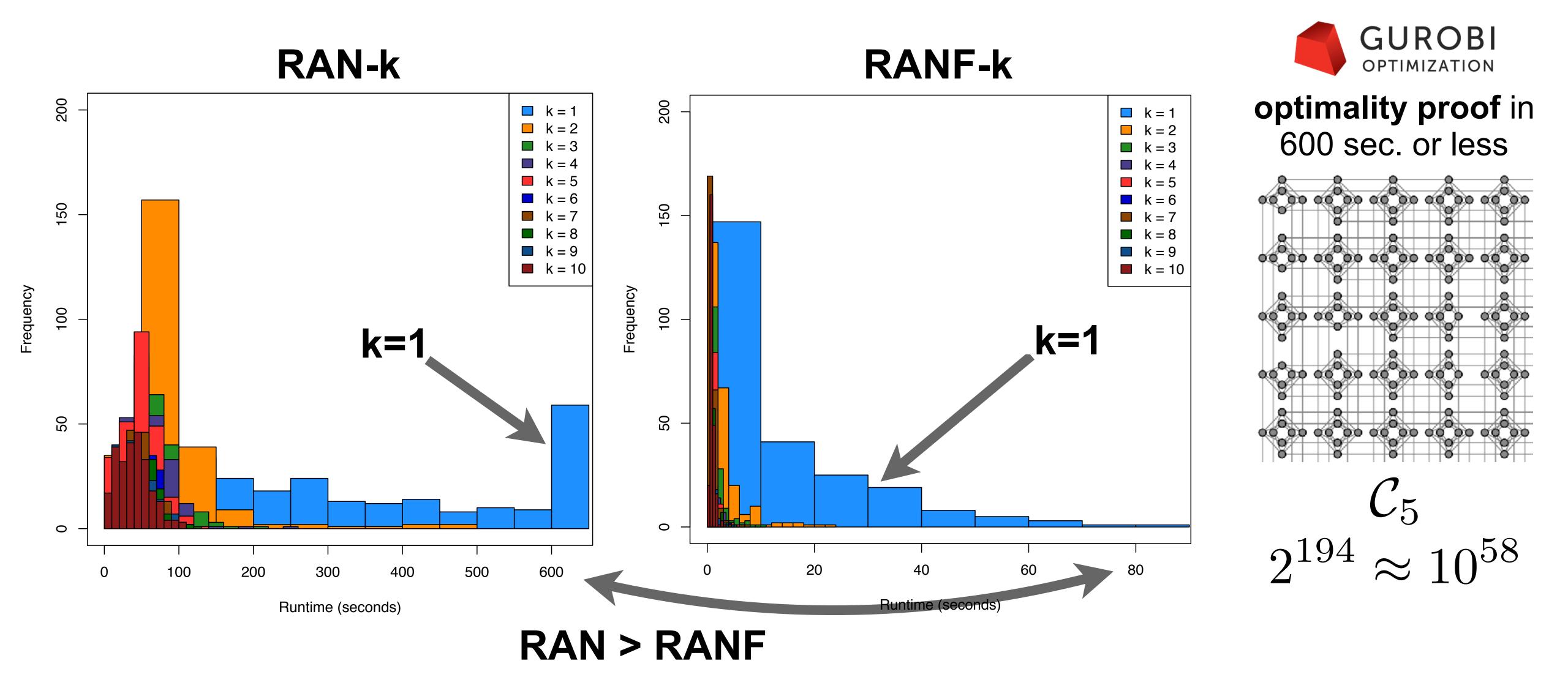
Too Easy Filtering



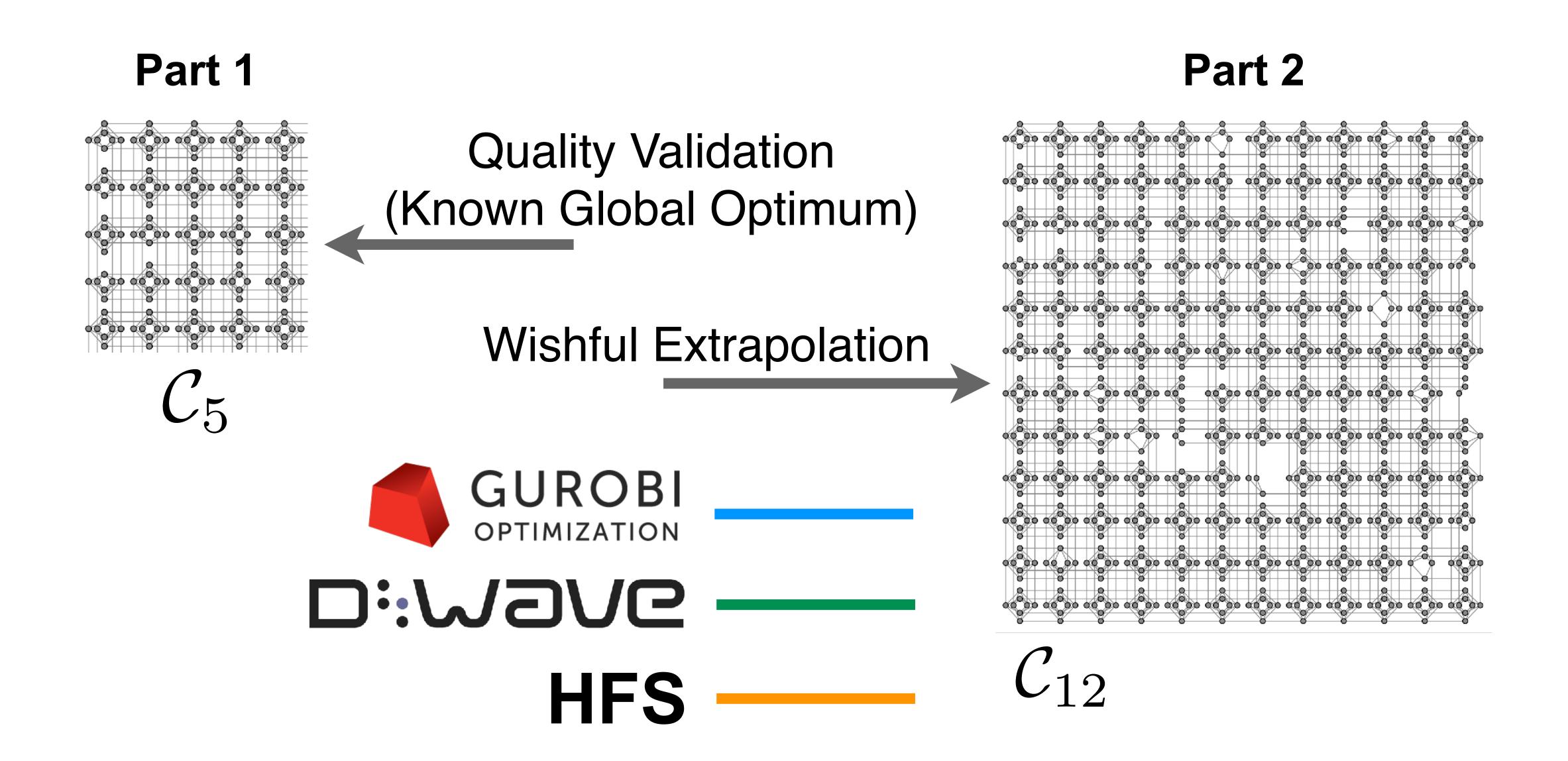
Problem Hardness



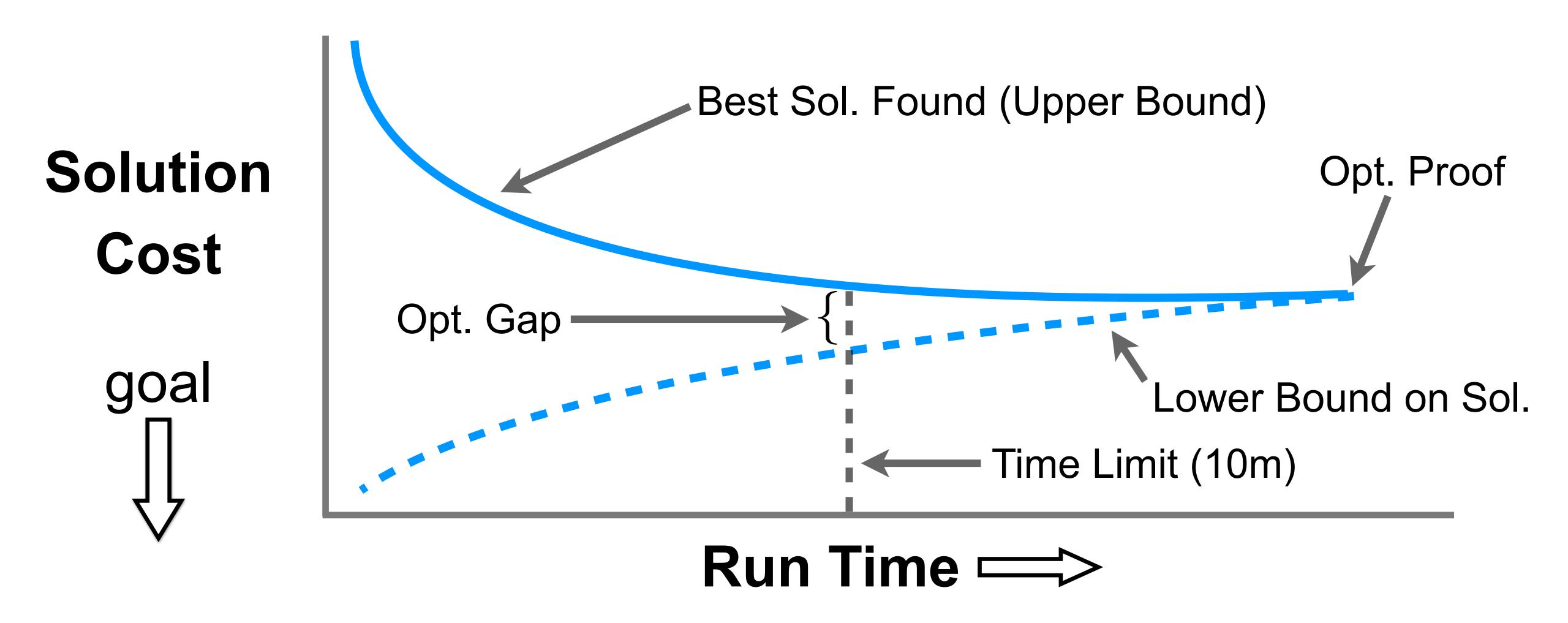
Problem Hardness



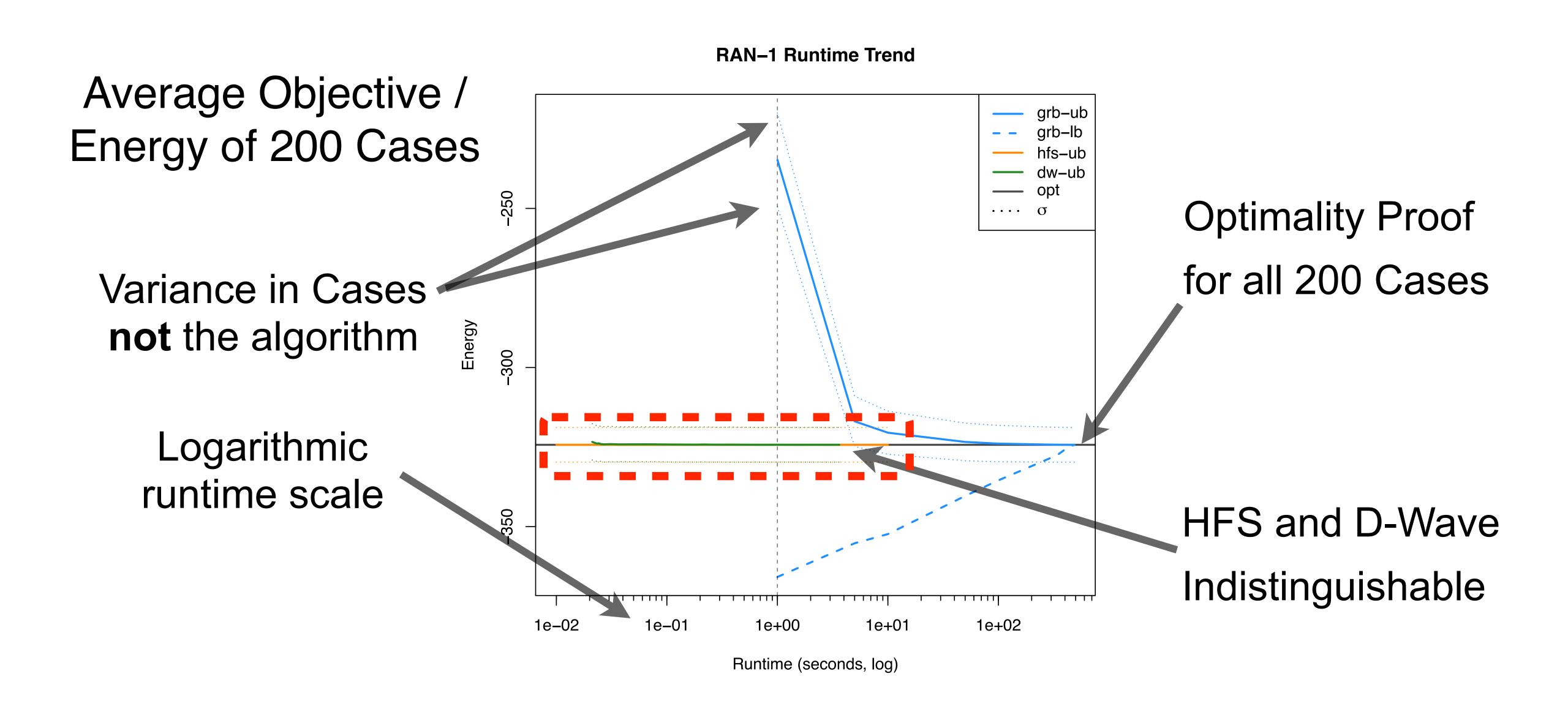
Detailed Benchmarking Study



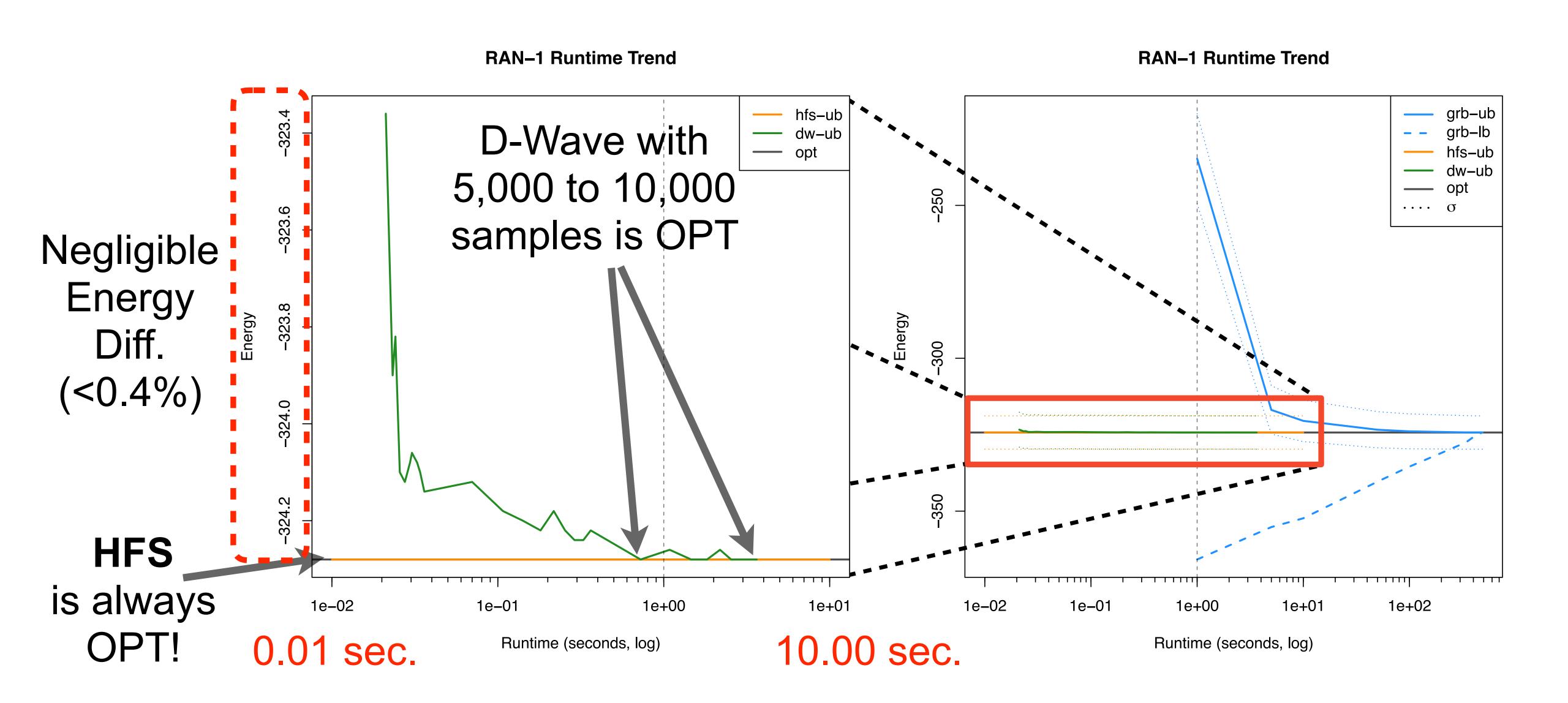
Detailed Benchmarking Study



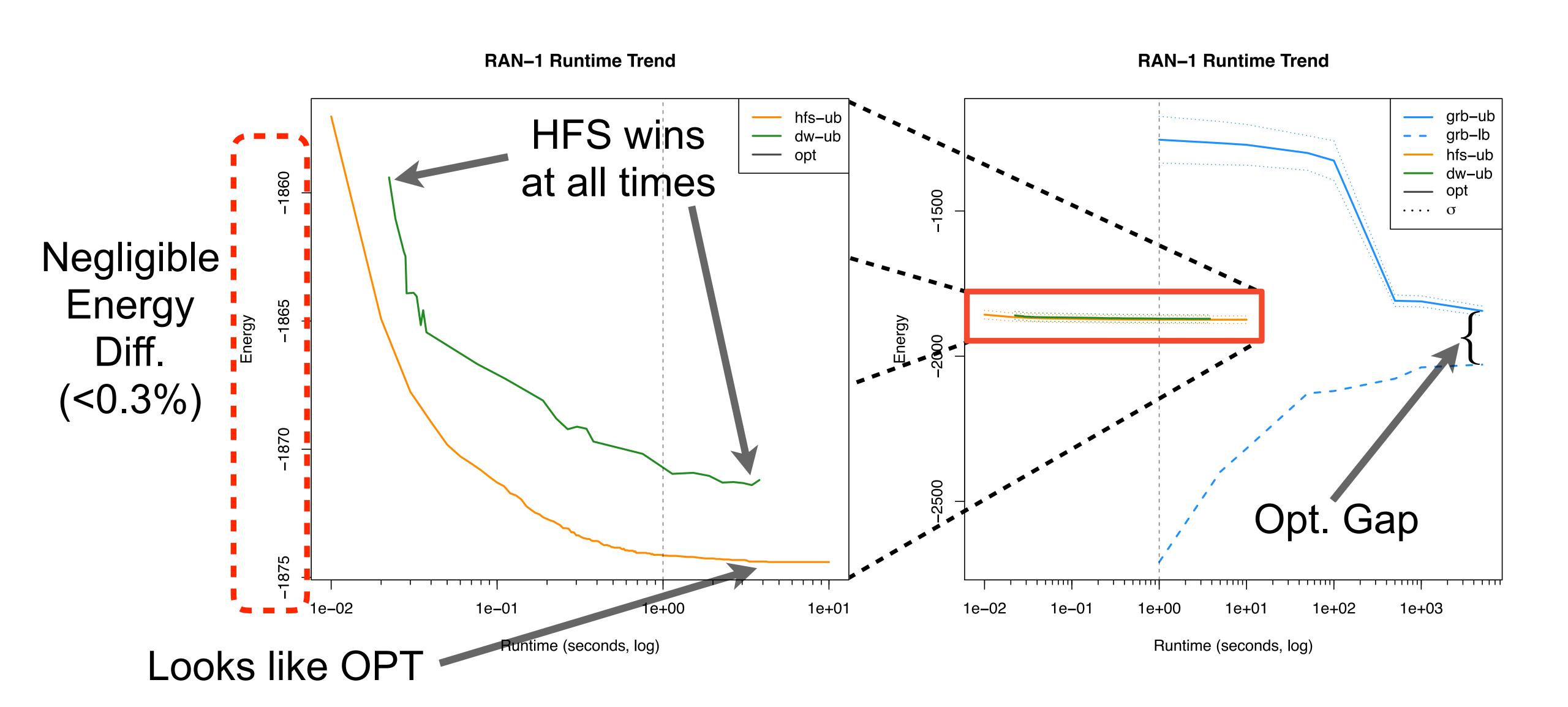
Detailed Benchmarking Study (RAN-1, C₅)



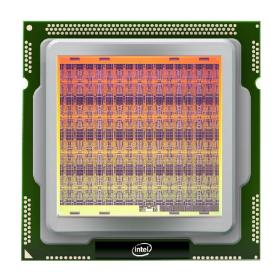
Detailed Benchmarking Study (RAN-1, C₅)



Detailed Benchmarking Study (RAN-1, C₁₂)

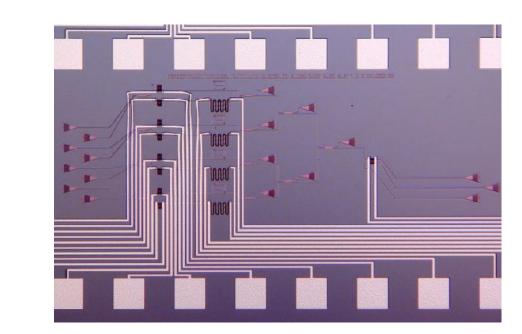


Concluding Thoughts



s.t.







QUBO

$$c_{ij}\sigma_i\sigma_j+\sum_i c_i$$

Ising

 $i{\in}\mathcal{N}$



$$\min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij} x_i x_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_i x_i \qquad \min: \sum_{i,j\in\mathcal{E}} \boldsymbol{c}_{ij} \sigma_i \sigma_j + \sum_{i\in\mathcal{N}} \boldsymbol{c}_i \sigma_i$$

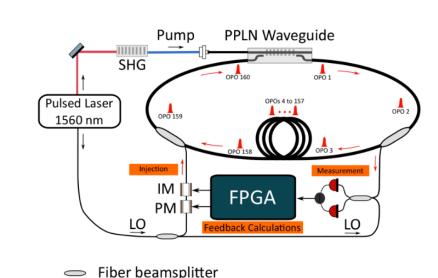
s.t.



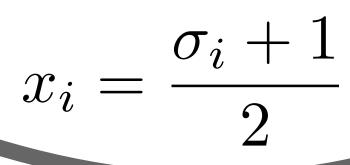
$$x_i \in \{0, 1\} \ \forall i \in \mathcal{N}$$

$$\sigma_i \in \{-1, 1\} \ \forall i \in \mathcal{N}$$

 $i,j \in \mathcal{E}$

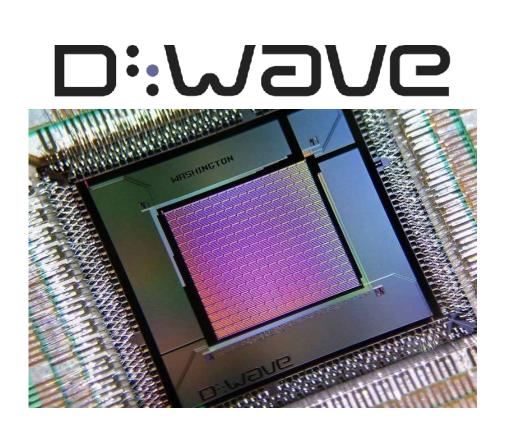






Ising Processing Units: Potential and Challenges for Discrete Optimization

https://arxiv.org/pdf/1707.00355.pdf



Concluding Thoughts

- Emerging computational hardware may fundamentally change optimization algorithm development
 - Ising Coprocessors are like ALUs or GPUs for optimization!
- Based on initial benchmarking efforts D-Wave hardware is not outperforming state-of-the-art alternatives, but seems competitive
 - I believe we are on the cusp of a performance breakthrough
- Watch out for other emerging hardware platforms!

Thanks!

Questions?